

# Design and Analysis of Algorithms

- *Course Code:* BCSE304L
- *Course Type:* Theory (ETH)
- *Slot:* A1+TA1 & & A2+TA2
- *Class ID:* VL2023240500901  
VL2023240500902

A1+TA1

Day	Start	End
Monday	08:00	08:50
Wednesday	09:00	09:50
Friday	10:00	10:50

A2+TA2

Day	Start	End
Monday	14:00	14:50
Wednesday	15:00	15:50
Friday	16:00	16:50

# Syllabus- Module 1

Module:1

Design Paradigms: Greedy, Divide and Conquer  
Techniques

6 hours

Overview and Importance of Algorithms - Stages of algorithm development: Describing the problem, Identifying a suitable technique, Design of an algorithm, Derive Time Complexity, Proof of Correctness of the algorithm, Illustration of Design Stages -

**Greedy techniques:** Fractional Knapsack Problem, and Huffman coding

**Divide and Conquer:** Maximum Subarray, Karatsuba faster integer multiplication algorithm.

# Maximum Subarray

## Definition:

- The Maximum Subarray problem involves finding the **contiguous** subarray with the largest sum within a given array of numbers.
- The problem of maximum subarray sum is basically finding the part of an array whose elements has the largest sum.

## Example:

- Let's consider an array  $arr = [-2, 1, -3, 4, -1, 2, 1, -5, 4]$ .
- The Maximum Subarray is  $[4, -1, 2, 1]$  with a sum of 6.

# Maximum Subarray

## Example:

array arr = [3,5,1,7,9].

The Maximum Subarray is [[3,5,1,7,9] with a max sum of  $3+5+1+7+9=25$ .

## Example:

array arr = [3,-1,-1,10,-3,-2,4].

The Maximum Subarray is [3,-1,-1,10] with a max sum of  $3-1-1+10=11$ .

# Maximum Subarray

## Time complexity in brute force approach:

A brute force approach would involve checking all possible subarrays and their sums, leading to a time complexity of  $O(n^2)$ , where  $n$  is the length of the array.

## Time complexity in Dive and Conquer approach:

The Divide and Conquer approach has a time complexity of  $O(n \log n)$ , where  $n$  is the length of the array.

**Recurrence relation:**  $T(n) = 2T(n/2) + O(n)$

# Maximum Subarray: brute force approach

Algorithm MaximumSubarrayBruteforce(arr):

maxsum = MIN\_INT      // Initialize maxsum to the smallest possible integer value

n = length of arr      // Get the length of the array

    // Outer loop: iterate through all possible starting indices

for i = 0 to n-1 do:

    sum = 0      // Initialize the sum for the current starting index

    // Inner loop: iterate through subarrays starting from index i

        for j = i to n-1 do:

            sum = sum + arr[j]      // Add the current element to the running sum

            // Update maxsum if the current sum is greater

            if sum >= maxsum then:

                maxsum = sum

return maxsum      // Return the overall maximum subarray sum

A brute force approach would involve checking all possible subarrays and their sums, leading to a time complexity of  $O(n^2)$ , where  $n$  is the length of the array.

# Maximum Subarray

Time complexity in **Dive and Conquer approach**:

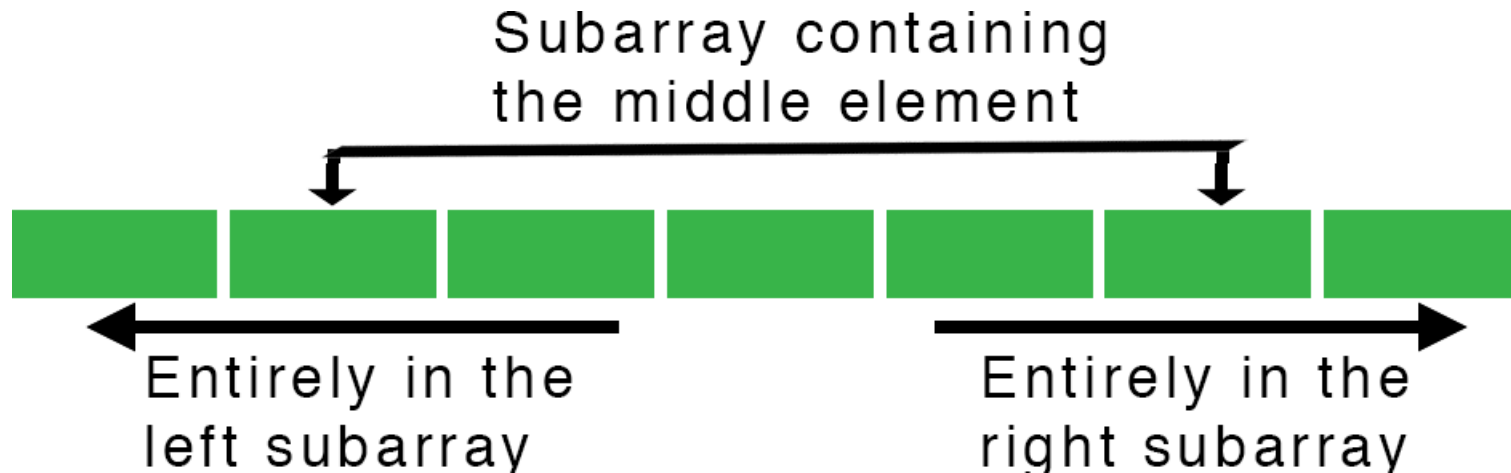
The Divide and Conquer approach has a time complexity of  $O(n \log n)$ , where  $n$  is the length of the array.

**Recurrence relation:**  $T(n) = 2T(n/2) + O(n)$

**Example:**

array arr = [3,-1,-1,10,-3,-2,4].

The Maximum Subarray is [3,-1,-1,10] with a max sum of  $3-1-1+10=11$ .

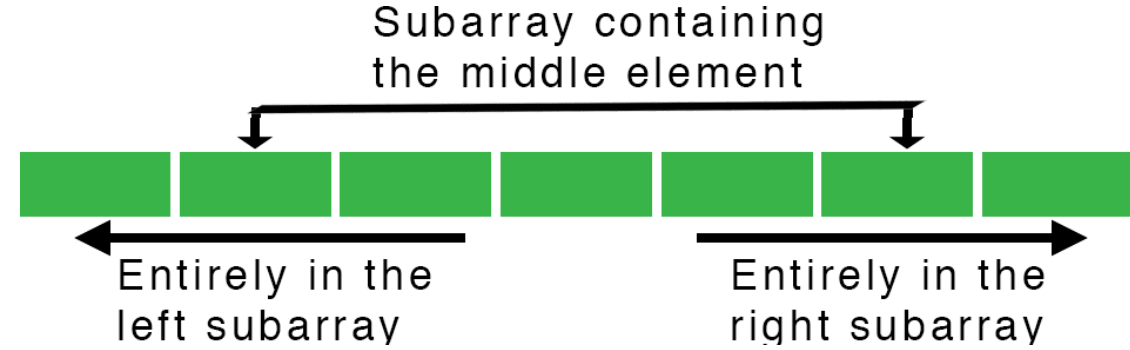
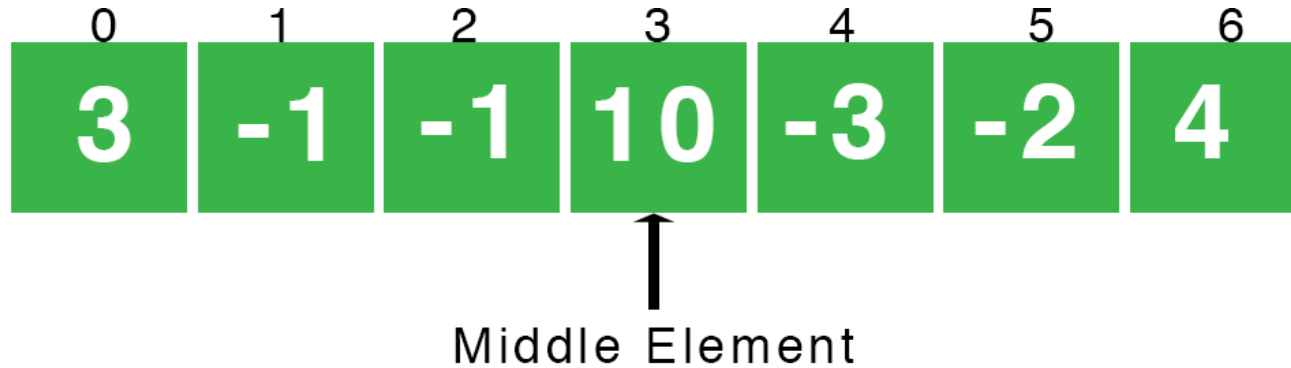


# Maximum Subarray

## Example:

array arr = [3,-1,-1,10,-3,-2,4].

The Maximum Subarray is [3,-1,-1,10] with a max sum of  $3-1-1+10=11$ .



Left subarray

sum = 0

Start from index 3:

$10 > 0 \Rightarrow \text{sum} = 10$

$10 - 1 = 9 < \text{sum}(10)$

$9 - 1 = 8 < \text{sum}(10)$

$8 + 3 = 11 > \text{sum} \Rightarrow \text{sum} = 11$

Right subarray

sum = 0

Start from index 4:

$-3 < \text{sum}(0)$ , sum is 0

$-3 - 2 = -5 < \text{sum}(0)$ , sum is 0

$-5 + 4 = -1 < \text{sum}(0)$ , sum is 0

**Final crossing sum =  $11 + 0 = 11$  from index 0 to 3**



# Maximum Subarray

```
function find_maximum_subarray(arr, low, high):  
    if low == high:  
        return low, high, arr[low]  
  
    mid = (low + high) / 2  
  
    left_low, left_high, left_sum = find_maximum_subarray(arr, low, mid)  
    right_low, right_high, right_sum = find_maximum_subarray(arr, mid + 1, high)  
    cross_low, cross_high, cross_sum = find_max_crossing_subarray(arr, low, mid, high)  
  
    if left_sum >= right_sum and left_sum >= cross_sum:  
        return left_low, left_high, left_sum  
    elif right_sum >= left_sum and right_sum >= cross_sum:  
        return right_low, right_high, right_sum  
    else:  
        return cross_low, cross_high, cross_sum
```

# Maximum Subarray

```
function find_max_crossing_subarray(arr, low, mid, high):  
    left_sum = negative infinity  
    sum = 0  
    max_left = 0  
  
    for i from mid downto low:  
        sum += arr[i]  
        if sum > left_sum:  
            left_sum = sum  
            max_left = i  
  
    right_sum = negative infinity  
    sum = 0  
    max_right = 0  
  
    for j from mid + 1 to high:  
        sum += arr[j]  
        if sum > right_sum:  
            right_sum = sum  
            max_right = j  
  
    return max_left, max_right, left_sum + right_sum
```

# Maximum Subarray

```
function find_max_crossing_subarray(arr, low, mid, high):
    left_sum = negative infinity
    sum = 0
    max_left = 0

    for i from mid downto low:
        sum += arr[i]
        if sum > left_sum:
            left_sum = sum
            max_left = i

    right_sum = negative infinity
    sum = 0
    max_right = 0

    for j from mid + 1 to high:
        sum += arr[j]
        if sum > right_sum:
            right_sum = sum
            max_right = j

    return max_left, max_right, left_sum + right_sum
```

```
function find_maximum_subarray(arr, low, high):
    if low == high:
        return low, high, arr[low]

    mid = (low + high) / 2

    left_low, left_high, left_sum = find_maximum_subarray(arr, low, mid)
    right_low, right_high, right_sum = find_maximum_subarray(arr, mid + 1, high)
    cross_low, cross_high, cross_sum = find_max_crossing_subarray(arr, low, mid, high)

    if left_sum >= right_sum and left_sum >= cross_sum:
        return left_low, left_high, left_sum
    elif right_sum >= left_sum and right_sum >= cross_sum:
        return right_low, right_high, right_sum
    else:
        return cross_low, cross_high, cross_sum
```

arr = [-2, 1, -3, 4, -1, 2, 1, -5, 4]

low, high, max\_sum = find\_maximum\_subarray(arr, 0, length(arr) - 1)

max\_subarray = arr[low:high + 1]

print("Maximum Subarray:", max\_subarray, "with sum", max\_sum)

# Maximum Subarray

Recurrence relation:  $T(n)=2T(n/2)+O(n)$

- This recurrence relation signifies that the problem is divided into two subproblems of size  $n/2$ , and the merging step (combining solutions) takes  $O(n)$  time.
- In the context of the Maximum Subarray problem, the recursive algorithm involves finding the maximum subarray in the left half, the right half, and a maximum subarray crossing the midpoint. The maximum of these three will be the solution for the entire array.

This can be represented as:

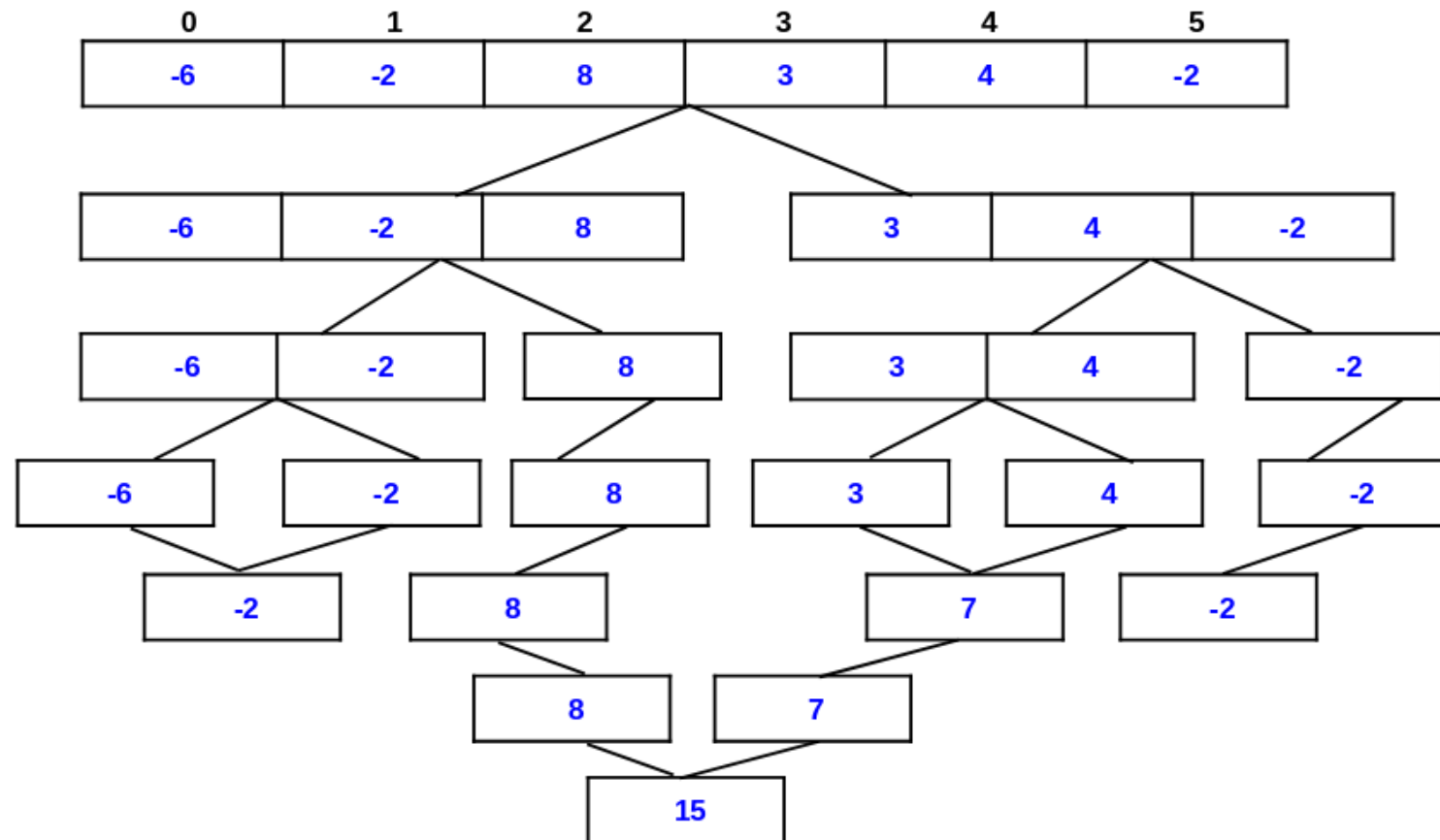
$\text{max\_subarray}(\text{arr}, \text{low}, \text{high}) = \max(\text{max\_subarray}(\text{arr}, \text{low}, \text{mid}), \text{max\_subarray}(\text{arr}, \text{mid}+1, \text{high}), \text{max\_crossing\_subarray}(\text{arr}, \text{low}, \text{mid}, \text{high}))$

Here,  $\text{max\_crossing\_subarray}(\text{arr}, \text{low}, \text{mid}, \text{high})$  finds the maximum subarray that crosses the midpoint.

# Maximum Subarray

Another example:

- Input array =  $[-6, -2, 8, 3, 4, -2]$ .
- here maximum contiguous subarray sum from left-side is 8.
- maximum contiguous subarray sum from right-side is 7.
- midpoint cross subarray sum is 15.
- hence the maximum subarray sum is 15.



# Karatsuba faster integer multiplication algorithm

Any

Question



PresenterMedia



# Thank You!

**FOR YOUR  
ATTENTION**

