

# CSE331 Introduction to Algorithms

## Lecture 5: Maximum Subarray

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# Introduction

- Last time, we saw a divide-and-conquer algorithm: MERGE SORT.
- Today, we will see another example of a divide-and-conquer algorithm.
- **Reference for this lecture:**
  - ▶ Section 4.1 of the textbook (p. 68–74)  
[Introduction to Algorithms](#) by Cormen, Leiserson, Rivest and Stein.

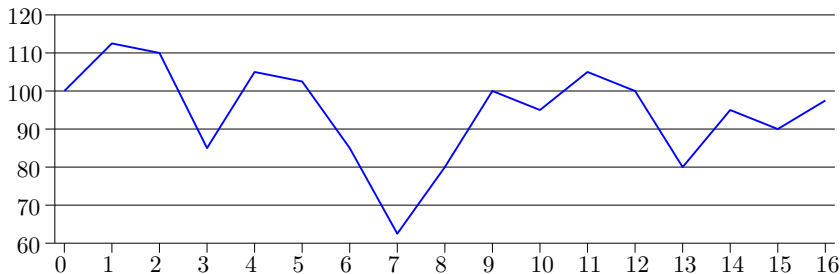
# The Divide-and-Conquer Approach

- MERGE SORT follows a *divide-and-conquer* approach.
- Divide and Conquer approach:
  - ▶ **Divide** the problem into a number of subproblems that are smaller instances of the same problem.
  - ▶ **Conquer** the subproblems by solving them recursively. If the subproblem sizes are small enough, however, just solve the subproblems in a straightforward manner.
  - ▶ **Combine** the solutions to the subproblems into the solution for the original problem.
- MERGE SORT follows this approach.
  - ▶ **Divide**: Divide the  $n$ -element sequence to be sorted into two subsequences of  $n/2$  elements each.
  - ▶ **Conquer**: Sort the two subsequences recursively using MERGE SORT.
  - ▶ **Combine**: Merge the two sorted subsequences to produce the sorted answer.

# The Divide-and-Conquer Approach

- The divide-and-conquer approach is one of the most important algorithm design technique.
- It can often be analyzed using the recursion tree method.
  - ▶ We will see other methods later this semester.
- Not all algorithms follow this approach.
- For instance, INSERTION SORT follows an *incremental* approach:
  - ▶ After sorting  $A[1 \dots j - 1]$ , we insert  $A[j]$  at the proper place.
  - ▶ An incremental algorithm adds elements one by one.
- In this lecture, we present a divide and conquer algorithm for the MAXIMUM SUBARRAY problem.
- We will see more examples in the next lectures.

# Buying and Selling

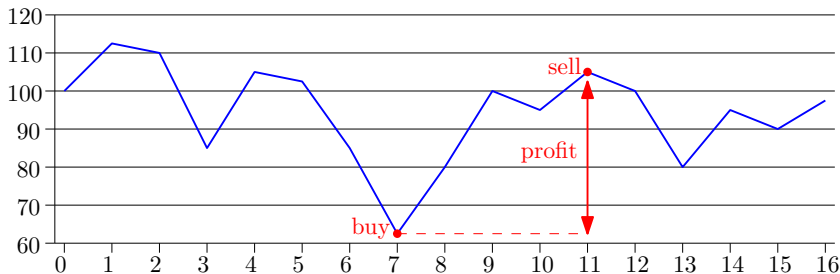


| day    | 0   | 1   | 2   | 3   | 4   | 5   | 6   | 7   | 8  | 9   | 10 | 11  | 12  | 13  | 14 | 15 | 16 |
|--------|-----|-----|-----|-----|-----|-----|-----|-----|----|-----|----|-----|-----|-----|----|----|----|
| price  | 100 | 113 | 110 | 85  | 105 | 102 | 86  | 63  | 81 | 101 | 94 | 106 | 101 | 79  | 94 | 90 | 97 |
| change |     | 13  | -3  | -25 | 20  | -3  | -16 | -23 | 18 | 20  | -7 | 12  | -5  | -22 | 15 | -4 | 7  |

## Problem

*When were the best times to buy and sell?*

# Buying and Selling



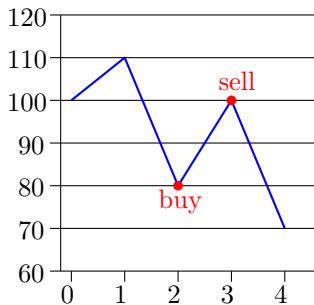
| day    | 0   | 1   | 2   | 3   | 4   | 5   | 6   | 7   | 8  | 9   | 10 | 11  | 12  | 13  | 14 | 15 | 16 |
|--------|-----|-----|-----|-----|-----|-----|-----|-----|----|-----|----|-----|-----|-----|----|----|----|
| price  | 100 | 113 | 110 | 85  | 105 | 102 | 86  | 63  | 81 | 101 | 94 | 106 | 101 | 79  | 94 | 90 | 97 |
| change |     | 13  | -3  | -25 | 20  | -3  | -16 | -23 | 18 | 20  | -7 | 12  | -5  | -22 | 15 | -4 | 7  |

## Answer

Buy on day 7, sell on day 11. Profit: 43

# Buying and Selling

- Difficulty: We don't necessarily buy at the lowest price or sell at the highest price.
- Example:





# Problem Transformation

- Consider the array  $A[1 \dots 16]$  of change in price.

|    |    |     |    |    |     |     |    |    |    |    |    |     |    |    |    |
|----|----|-----|----|----|-----|-----|----|----|----|----|----|-----|----|----|----|
| 1  | 2  | 3   | 4  | 5  | 6   | 7   | 8  | 9  | 10 | 11 | 12 | 13  | 14 | 15 | 16 |
| 13 | -3 | -25 | 20 | -3 | -16 | -23 | 18 | 20 | -7 | 12 | -5 | -22 | 15 | -4 | 7  |

Maximum subarray  $A[8 \dots 11]$

- The *maximum subarray* is the contiguous subarray whose elements have the largest sum. Here, it is  $A[8 \dots 11]$ .
- So the best times to buy and sell are days 7 and 11.
- We reduced our original problem to:

## Problem

Given an array  $A[1 \dots n]$  of  $n$  numbers, find the indices  $p, q$  such that  $1 \leq p \leq q \leq n$  and  $\sum_{i=p}^q A[i]$  is maximum.

# The MAXIMUM SUBARRAY Problem

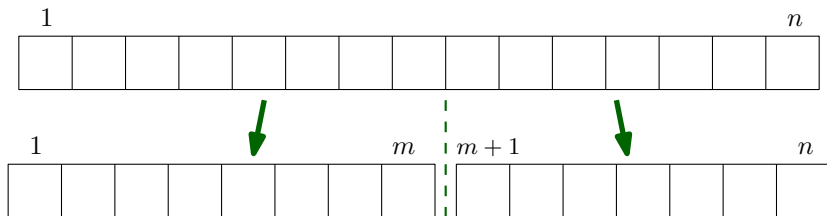
## Brute-force approach to MAXIMUM SUBARRAY

```
1: procedure MAXIMUM-SUBARRAY( $A, 1, n$ )
2:    $\text{max} \leftarrow -\infty$ 
3:   for  $i \leftarrow 1, n$  do
4:     for  $j \leftarrow i, n$  do
5:        $\text{sum} \leftarrow 0$ 
6:       for  $k \leftarrow i, j$  do
7:          $\text{sum} \leftarrow \text{sum} + A[k]$ 
8:       if  $\text{sum} > \text{max}$  then
9:          $\text{max} \leftarrow \text{sum}, p \leftarrow i, q \leftarrow j$ 
10:  return  $p, q, \text{max}$ 
```

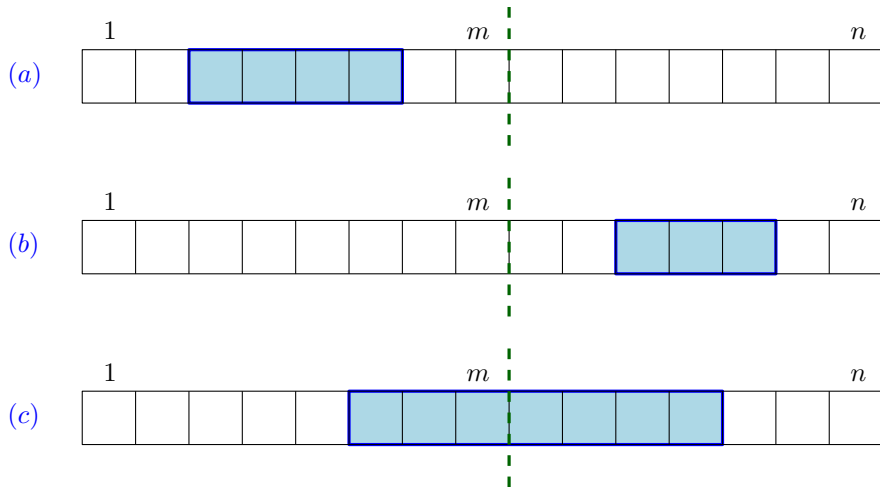
- Running time?  $\Theta(n^3)$ .
- Proof done in class. See lecture notes.

# Divide-and-Conquer Approach

- This is not very good, because our original problem of buying and selling stock could be solved by brute force in time  $\Theta(n^2)$ .
- We will now give a  $\Theta(n \log n)$ -time divide and conquer algorithm.
- (It is possible to solve MAXIMUM SUBARRAY in linear time. See [Wikipedia](#).)
- Divide & conquer approach: *Divide*  $A[1 \dots n]$  into  $A[1 \dots m]$ ,  $A[m + 1 \dots n]$  where  $m = \lceil n/2 \rceil$ .



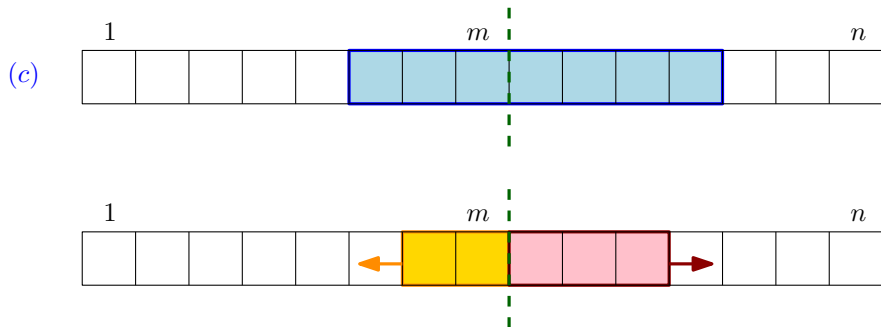
# The MAXIMUM SUBARRAY Problem



# The MAXIMUM SUBARRAY Problem

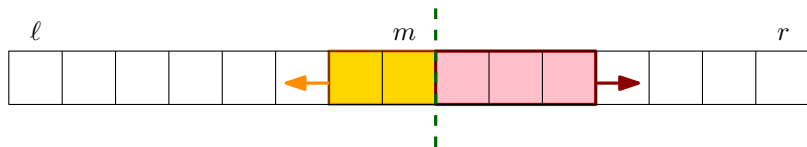
- Three cases: The maximum subarray
  - (a) is contained in  $A[1 \dots m]$ ,
  - (b) is contained in  $A[m + 1 \dots n]$ , or
  - (c) crosses the boundary between the two subarrays.
- Cases (a) and (b) are handled by recursing on  $A[1 \dots m]$  and  $A[m + 1 \dots n]$ , respectively. This is the *conquer* step.
- Case (c) is non-trivial. We will give an efficient algorithm for it, which is the main part of the *combine* step.

# Maximum Crossing Subarray



- Idea: we can maximize the left part and the right part separately, and then combine the two maxima.

# Maximum Crossing Subarray



- Similarly as we did for MERGE SORT, our algorithm solves this problem for the subarray  $A[\ell \dots r]$  split at index  $m = \lfloor (\ell + r)/2 \rfloor$ .

# Maximum Crossing Subarray

## Pseudocode

```
1: procedure MAX-CROSSING-SUBARRAY( $A, \ell, m, r$ )
2:    $M_1 \leftarrow -\infty$ ,  $\text{sum} \leftarrow 0$ 
3:   for  $i \leftarrow m + 1, r$  do
4:      $\text{sum} \leftarrow \text{sum} + A[i]$ 
5:     if  $\text{sum} > M_1$  then
6:        $M_1 \leftarrow \text{sum}$ ,  $q \leftarrow i$ 
7:    $M_2 \leftarrow -\infty$ ,  $\text{sum} \leftarrow 0$ 
8:   for  $i \leftarrow m, \ell$  do
9:      $\text{sum} \leftarrow \text{sum} + A[i]$ 
10:    if  $\text{sum} > M_2$  then
11:       $M_2 \leftarrow \text{sum}$ ,  $p \leftarrow i$ 
12:   return  $p, q, M_1 + M_2$ 
```

▷  $i$  goes down from  $m$  to  $\ell$

- The result is  $A[p \dots q]$  with sum  $M_1 + M_2$ .



# Maximum Crossing Subarray

- Analysis: runs in *linear time*.
- More precisely,  $\Theta(n')$  where  $n' = r - \ell + 1$  is the size of the input subarray  $A[\ell \dots r]$ .

# The MAXIMUM SUBARRAY Problem

## Pseudocode

```
1: procedure MAXIMUM-SUBARRAY( $A, \ell, r$ )
2:   if  $\ell = r$  then
3:     return  $\ell, r, A[\ell]$                                 ▷ base case: 1-element array
4:    $m \leftarrow \lfloor (\ell + r)/2 \rfloor$ 
5:    $p_a, q_a, M_a \leftarrow \text{MAXIMUM-SUBARRAY}(A, \ell, m)$       ▷ case (a)
6:    $p_b, q_b, M_b \leftarrow \text{MAXIMUM-SUBARRAY}(A, m + 1, r)$     ▷ case (b)
7:    $p_c, q_c, M_c \leftarrow \text{MAX-CROSSING-SUBARRAY}(A, \ell, m, r)$  ▷ case (c)
8:    $r = \arg \max_{x \in \{a, b, c\}} M_x$                             ▷ i.e.  $M_r = \max(M_a, M_b, M_c)$ 
9:   return  $p_r, q_r, M_r$ 
```

# The MAXIMUM SUBARRAY Problem

- **Divide:** Line 4
- **Conquer:** Lines 5–6
- **Combine:** Lines 7–9
- Analysis:
  - ▶ Suppose  $n$  is a power of 2, and  $n \geq 2$ . ( $n = 2^h$  for some integer  $h \geq 1$ .)
  - ▶ let  $T(n)$  be the running time.
  - ▶ Lines 5 and 6 take  $T(n/2)$  each.
  - ▶ Line 7 takes  $\Theta(n)$ .
  - ▶ Other lines take  $\Theta(1)$ .
  - ▶ Therefore

$$T(n) = 2T(n/2) + \Theta(n)$$

# The MAXIMUM SUBARRAY Problem

- Combined with the base case (Line 2–3) we obtain the recurrence relation:

$$T(n) = \begin{cases} \Theta(1) & \text{if } n = 1 \\ 2T(n/2) + \Theta(n) & \text{if } n \geq 2 \end{cases}$$

- $T(n) = \Theta(n \log n)$  because we found the same relation for MERGE SORT.