

Divide-and-Conquer

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Outline

- **Introduction**
- The Maximum-Subarray Problem
- The Matrix Multiplication Problem

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Divide-and-Conquer

- **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.

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Recurrences (1/2)

- A **recurrence** is an equation or inequality that describes a function in terms of its value on smaller inputs.
 - The worst-case running time $T(n)$ of the **MERGE-SORT** procedure can be described by the recurrence

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

whose solution we claimed to be $T(n) = \Theta(n \lg n)$.

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Recurrences (2/2)

- Recurrences can take many forms:
 - A recursive algorithm might divide subproblems into unequal sizes, such as a 2/3-to-1/3 split.
 - $T(n) = T(2n/3) + T(n/3) + \Theta(n)$.
 - Subproblems are not necessarily constrained to being a constant fraction of the original problem size.
 - For example, a recursive version of linear search would create just one subproblem containing only one element fewer than the original problem.
 - $T(n) = T(n - 1) + \Theta(1)$.

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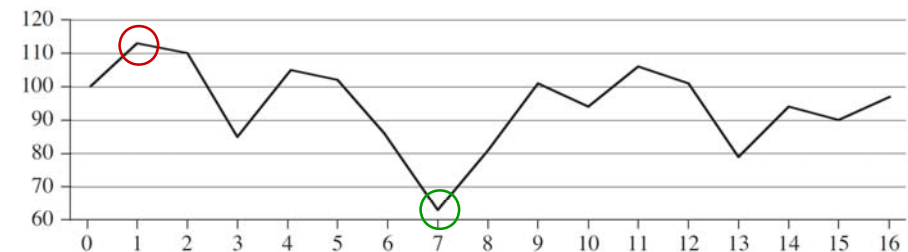
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Maximum-Subarray Problem (1/2)

- Suppose that you been offered the opportunity to invest a company. The stock price of the company is rather volatile.
- You are allowed to buy one unit of stock only one time and then sell it at a later date, buying and selling after the close of trading for the day.
- To compensate for this restriction, you are allowed to learn what the price of the stock will be in the future. Your goal is to **maximize your profit**.

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Maximum-Subarray Problem (2/2)



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

- You would want to buy at the lowest possible price and later on sell at the highest possible price—to maximize your profit.
- ☞ Unfortunately, you might not be able to buy at the lowest price and then sell at the highest price within a given period.

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How to Maximize Your Profit? (1/2)

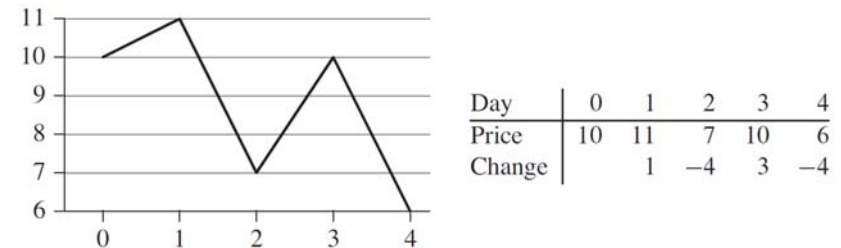
- You might think that you can always maximize profit by either buying at the lowest price or selling at the highest price.



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How to Maximize Your Profit? (2/2)

- How about this case?



- ➔ The maximum profit does not always start at the lowest price or end at the highest price!

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Brute-Force Solution

- We can easily devise a brute-force solution to this problem: just try every possible pair of buy and sell dates in which the buy date precedes the sell date.
- ➔ A period of n days has $C(n, 2)$ such pairs of dates, and this approach would take $\Omega(n^2)$ time.

💡 Can we do better?

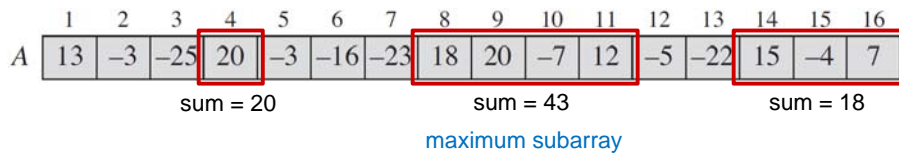
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Transformation

- We want to find a sequence of days over which the net change from the first day to the last is maximum.
- Instead of looking at the daily prices, let us instead consider the **daily change in price**, where the change on day i is the difference between the prices after day $i - 1$ and after day i .
- We now want to find the nonempty, contiguous subarray of A with the largest sum.
 - We call this contiguous subarray the **maximum subarray**.

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Maximum Subarray

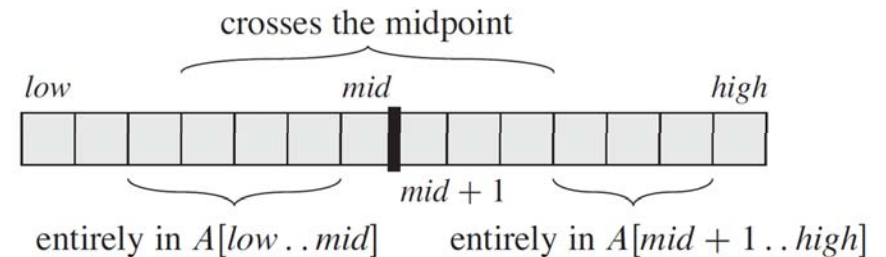


- The maximum-subarray problem is interesting only when the array contains some **negative numbers**.
 - If all the array entries were nonnegative, then the maximum-subarray problem would present no challenge, since the entire array would give the greatest sum.

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Solution Using Divide-and-Conquer (1/2)

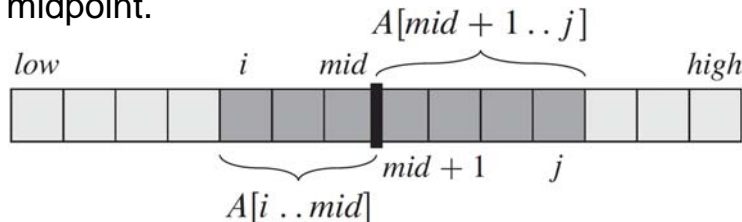
- Suppose we want to find a maximum subarray of the subarray $A[\text{low} \dots \text{High}]$.
- Divide-and-conquer suggests that we divide the subarray into two subarrays of as equal size as possible.



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Solution Using Divide-and-Conquer (2/2)

- To find a maximum subarray crossing the midpoint is not a smaller instance of our original problem, because it has the added restriction that the subarray it chooses must cross the midpoint.



- We just need to find maximum subarrays of the form $A[i \dots \text{mid}]$ and $A[\text{mid} + 1 \dots j]$ and then combine them.

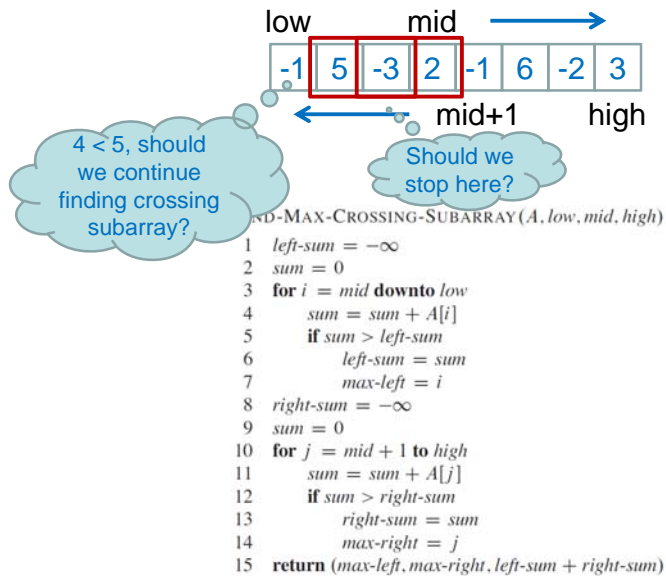
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FIND-MAX-CROSSING-SUBARRAY ($A, \text{low}, \text{mid}, \text{high}$)

```

1  left-sum =  $-\infty$ 
2  sum = 0
3  for i = mid downto low
4      sum = sum + A[i]
5      if sum > left-sum
6          left-sum = sum
7          max-left = i
8  right-sum =  $-\infty$ 
9  sum = 0
10 for j = mid + 1 to high
11     sum = sum + A[j]
12     if sum > right-sum
13         right-sum = sum
14         max-right = j
15 return (max-left, max-right, left-sum + right-sum) 16
    
```

Illustration



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FIND-MAXIMUM-SUBARRAY(*A*, *low*, *high*)

```

1  if high == low
2      return (low, high, A[low]) // base case: only one element
3  else mid = ⌊(low + high)/2⌋
4      (left-low, left-high, left-sum) =
5          FIND-MAXIMUM-SUBARRAY(A, low, mid)
6      (right-low, right-high, right-sum) =
7          FIND-MAXIMUM-SUBARRAY(A, mid + 1, high)
8      (cross-low, cross-high, cross-sum) =
9          FIND-MAX-CROSSING-SUBARRAY(A, low, mid, high)
10     if left-sum ≥ right-sum and left-sum ≥ cross-sum
11         return (left-low, left-high, left-sum)
12     elseif right-sum ≥ left-sum and right-sum ≥ cross-sum
13         return (right-low, right-high, right-sum)
14     else return (cross-low, cross-high, cross-sum)
    
```

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The Matrix Multiplication Problem

- If $A = (a_{ij})$ and $B = (b_{ij})$ are square $n \times n$ matrices, then in the product $C = A \cdot B$, we define the entry

$$c_{ij}, \text{ for } i, j = 1, 2, \dots, n, \text{ by } c_{ij} = \sum_{k=1}^n a_{ik} \cdot b_{kj}.$$

SQUARE-MATRIX-MULTIPLY(*A*, *B*)

```

1  n = A.rows
2  let C be a new n × n matrix
3  for i = 1 to n
4      for j = 1 to n
5          cij = 0
6          for k = 1 to n
7              cij = cij + aik · bkj
8  return C
    
```

It takes $\Theta(n^3)$ time.

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