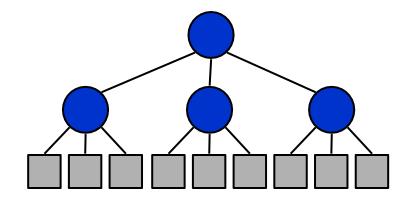
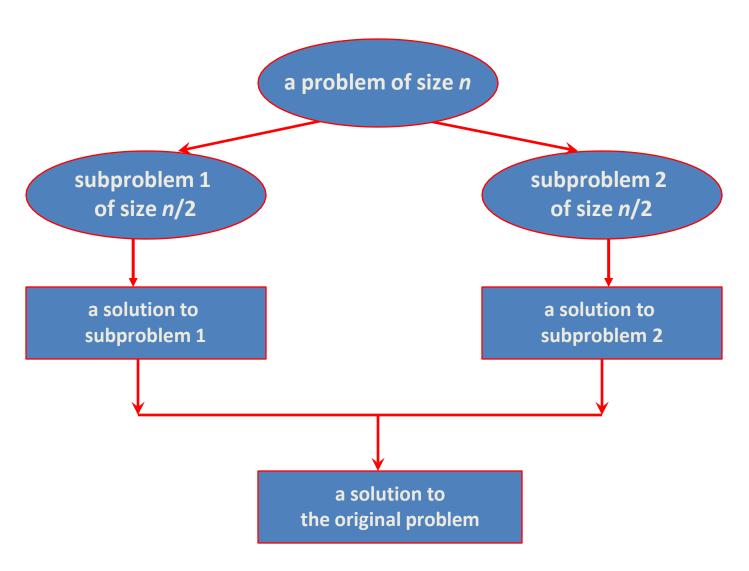
Divide-and-Conquer Technique: Maximum Subarray problem

Divide-and-Conquer

- Divide-and-Conquer is a general algorithm design paradigm:
 - Divide the problem into a number of subproblems that are smaller instances of the same problem
 - Conquer the subproblems by solving them recursively
 - Combine the solutions to the subproblems into the solution for the original problem
- ☐ The base case for the recursion are subproblems of constant size
- Analysis can be done using recurrence equations

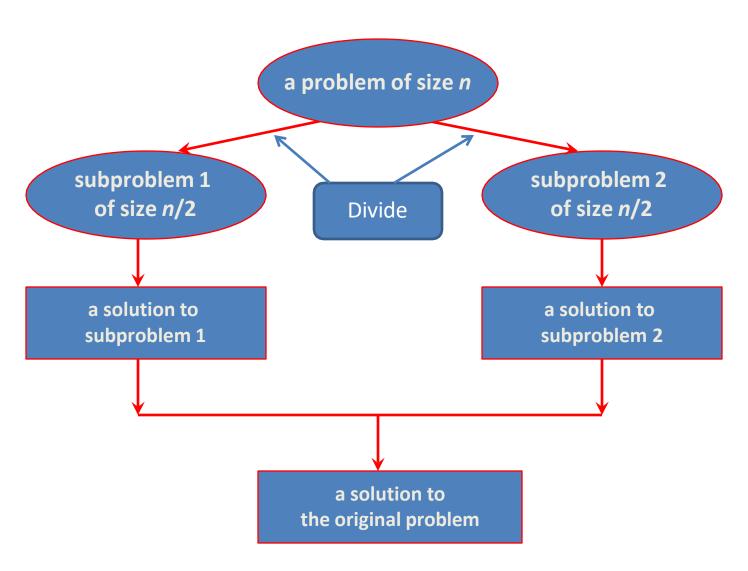


Divide-and-Conquer



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Maximum Subarray Problem

- *Input*: an array A[1..n] of n numbers
 - Assume that some of the numbers are negative,
 because this problem is trivial when all numbers are nonnegative
- *Output:* a nonempty subarray A[i..j] having the largest sum $S[i,j] = a_i + a_{i+1} + ... + a_i$

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

Maximum Subarray Problem

- *Input*: an array A[1..n] of n numbers
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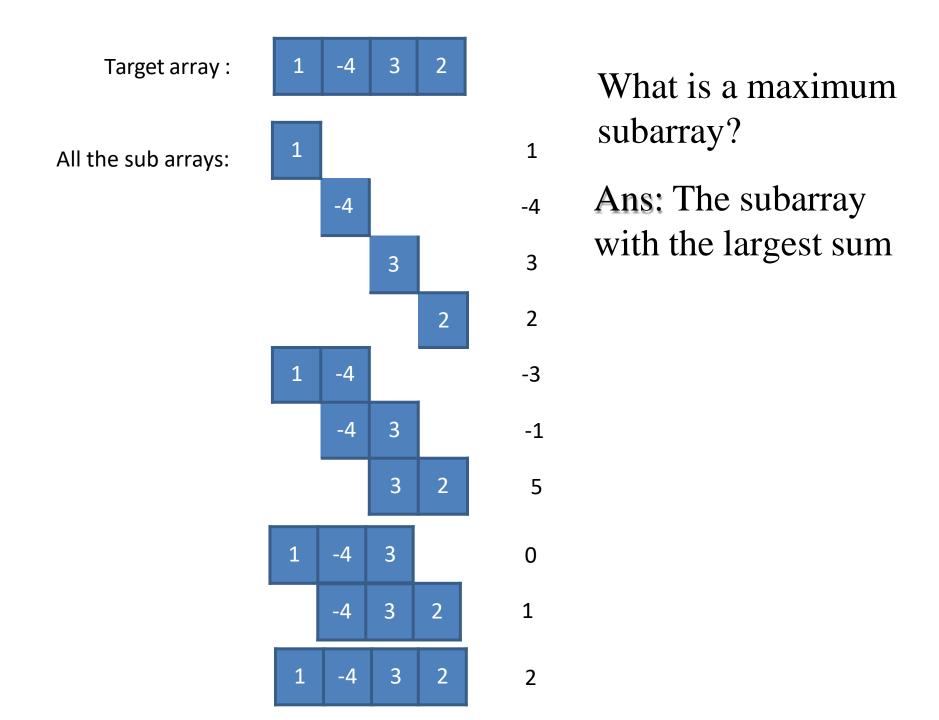
maximum subarray

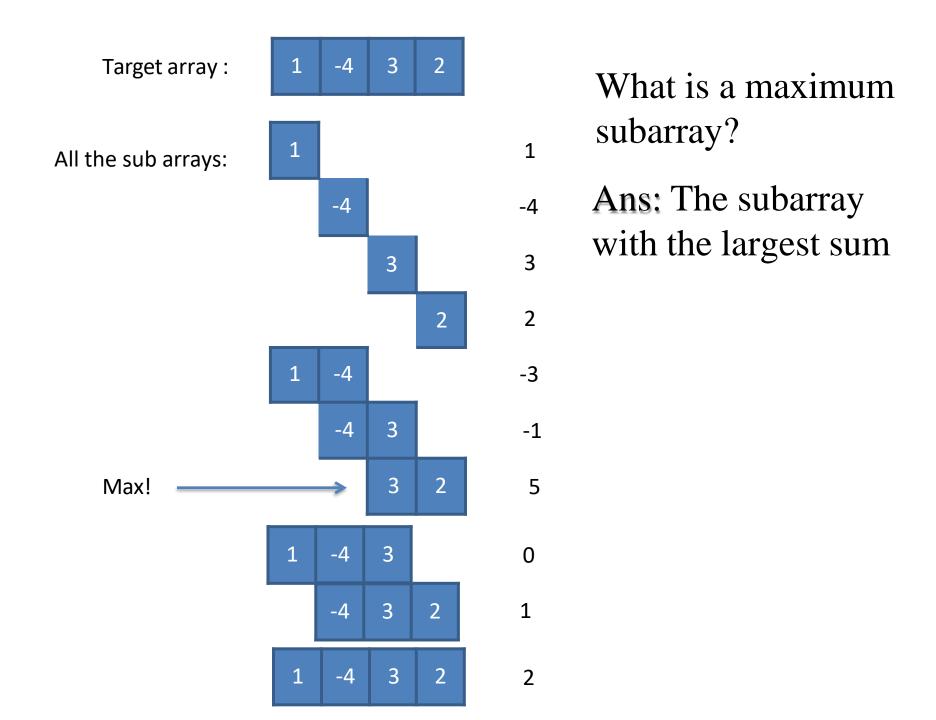
Target array :

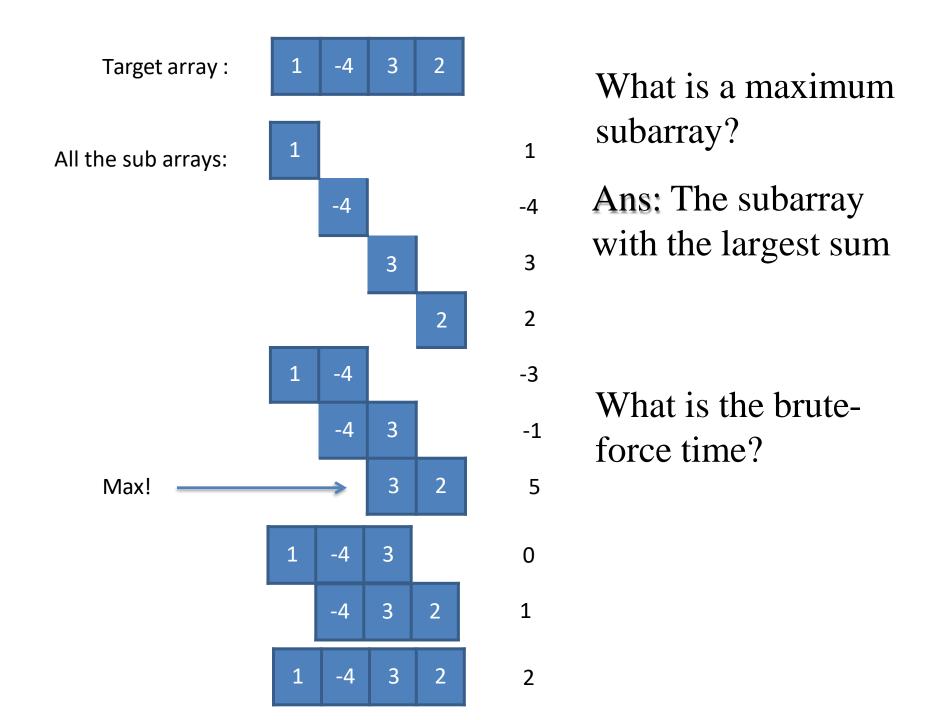
1 -4 3 2

All the sub arrays:

What is a maximum subarray?







All possible contiguous subarrays

```
A[1..1], A[1..2], A[1..3], ..., A[1..(n-1)], A[1..n]
```

$$\triangle$$
 A[2..2], A[2..3], ..., A[2..(n -1)], A[2.. n]

...

A[(n-1)..(n-1)], A[(n-1)..n]

A[n..n]

All possible contiguous subarrays

```
\triangle A[1..1], A[1..2], A[1..3], ..., A[1..(n-1)], A[1..n]
```

$$\triangle$$
 A[2..2], A[2..3], ..., A[2..(n -1)], A[2.. n]

□ ...

A[(n-1)..(n-1)], A[(n-1)..n]

A[n..n]

How many of them in total?

All possible contiguous subarrays

```
\triangle A[1..1], A[1..2], A[1..3], ..., A[1..(n-1)], A[1..n]
```

$$\triangle$$
 A[2..2], A[2..3], ..., A[2..(n -1)], A[2.. n]

□ ...

A[(n-1)..(n-1)], A[(n-1)..n]

A[n..n]

How many of them in total? $\circ \circ \circ \subset O(n^2)$

All possible contiguous subarrays

- \triangle A[1..1], A[1..2], A[1..3], ..., A[1..(n-1)], A[1..n]
- A[2..2], A[2..3], ..., A[2..(n-1)], A[2..n]
- ...
- A[(n-1)..(n-1)], A[(n-1)..n]
- A[n..n]

How many of them in total? $\circ \circ \circ \subset O(n^2)$

Algorithm: For each subarray, compute the sum.

Find the subarray that has the maximum sum.

Example:	2	-6	-1	3	-1	2	-2
sum from A[1]:	2	-4	-5	-2	-3	-1	-3
sum from A[2]:		-6	-7	-4	-5	-3	-5
sum from A[3]:			-1	2	1	3	1
sum from A[4]:				3	2	$\left(4\right)$	2
sum from A[5]:					-1	1	-1
sum from A[6]:						2	0
sum from A[7]:							-2

Outer loop: index variable i to indicate start of subarray, for $1 \le i \le n$, i.e., A[1], A[2], ..., A[n]

of for i = 1 to n do ...

Inner loop: for each start index i, we need to go through A[i..i], A[i..(i+1)], ..., A[i..n]

- use an index j for $i \le j \le n$, i.e., consider A[i..j]
- for j = i to n do ...

```
\max = -\infty
for i = 1 to n do
begin
    sum = 0
    for j = i to n do
    begin
        sum = sum + A[j]
        if sum > max
        then max = sum
    end
end
```

```
\max = -\infty
                                     Time
for i = 1 to n do
                                 complexity?
begin
                                    O(n^2)
    sum = 0
    for j = i to n do
    begin
        sum = sum + A[j]
        if sum > max
        then max = sum
    end
end
```

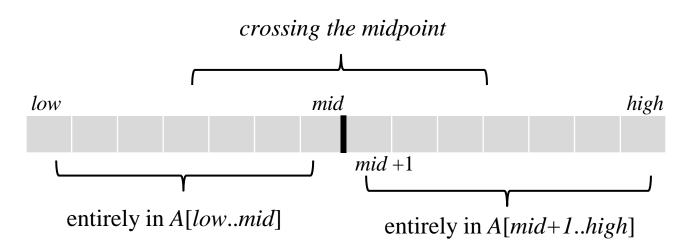
Possible locations of a maximum subarray A[i..j] of A[low..high], where $mid = \lfloor (low + high)/2 \rfloor$

• entirely in A[low..mid]

- $(low \le i \le j \le mid)$
- entirely in *A*[*mid*+1..*high*]
- $(mid < i \le j \le high)$

crossing the midpoint

 $(low \le i \le mid < j \le high)$

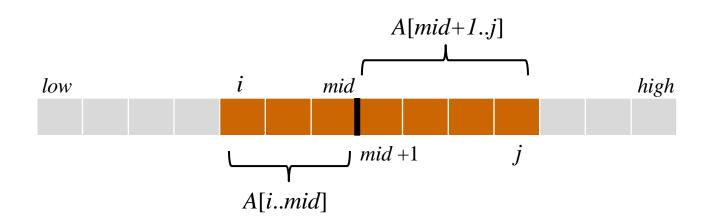


Possible locations of subarrays of A[low..high]

```
FIND-MAXIMUM-SUBARRAY (A, low, high)
  if high == low
      return (low, high, A[low]) // base case: only one element
  else mid = \lfloor low + high/2 \rfloor
      (left-low, left-high, left-sum) =
             FIND-MAXIMUM-SUBARRAY(A, low, mid)
      (right-low, right-high, right-sum) =
            FIND-MAXIMUM-SUBARRAY(A, mid + 1, high)
      (cross-low, cross-high, cross-sum) =
           FIND-MAX-CROSSING-SUBARRAY(A, low, mid, high)
       if left-sum \ge right-sum and left-sum \ge cross-sum
            return (left-low, left-high, left-sum)
       elseif right-sum \ge left-sum and right-sum \ge cross-sum
             return (right-low, right-high, right-sum)
       else return (cross-low, cross-high, cross-sum)
Initial call: FIND-MAXIMUM-SUBARRAY (A, 1, n)
```

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```
FIND-MAX-CROSSING-SUBARRAY (A, low, mid, high)
   left-sum = -\infty // Find a maximum subarray of the form A[i.mid]
   sum = 0
   for i = mid downto low
      sum = sum + A[i]
      if sum > left-sum
       left-sum = sum
        max-left = i
   right-sum = -\infty // Find a maximum subarray of the form A[mid + 1 ... j]
   sum = 0
   for j = mid + 1 to high
      sum = sum + A[j]
      if sum > right-sum
        right-sum = sum
        max-right = j
   // Return the indices and the sum of the two subarrays
   return (max-left, max-right, left-sum + right-sum)
```



A[i..j] comprises two subarrays A[i..mid] and A[mid+1..j]

 \Rightarrow maximum subarray crossing *mid* is S[4..9] = 16

Analyzing time complexity

FIND-MAX-CROSSING-SUBARRAY :
$$\Theta(n)$$
, where $n = high - low + 1$

FIND-MAXIMUM-SUBARRAY

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

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

= $\Theta(n \lg n)$ (similar to merge-sort)

Conclusion: Divide-and-Conquer

- This Divide and conquer algorithm is clearly substantially faster than any of the brute-force methods. It required some cleverness, and the programming is a little more complicated but the payoff is large.
- Divide and conquer is just one of several powerful techniques for algorithm design
- Divide-and-conquer algorithms can be analyzed using recurrences
- Can lead to more efficient algorithms