CSE 421 Section 4

Divide and Conquer

Administrivia



Announcements & Reminders

- HW2
 - If you think something was graded incorrectly, submit a regrade request!
- HW3
 - \circ Was due yesterday, 10/18
 - Remember, this quarter we have a LATE PROBLEM DAYS policy, instead of a late assignments policy
 - Total of up to **10 late problem days**
 - At most **2 late days per problem**
- HW4
 - Due Wednesday 10/25 @ 11:59pm

Writing a Divide and Conquer Algo



Divide and Conquer

- 1. Divide instance into subparts
- 2. Solve the parts recursively
- **3. Conquer** by combining the answers

The keys to this strategy:

- Come up with a baseline!
- Once you have your algo, write a recurrence for the runtime
 - Your d&c runtime should be BETTER than the baseline runtime

The Strategy (hint: it's the same as last week!)

- 1. Read and Understand the Problem
- 2. Generate Examples
- 3. Produce a Baseline
- 4. Brainstorm and Analyze Possible Algorithms
- 5. Write an Algorithm
- 6. Show Your Algorithm is Correct
- 7. Optimize and Analyze the Run Time

Input: An array of ints (possibly both positive and negative) **Output:** The largest possible sum of a (contiguous) subarray $A[i] + A[i + 1] + \dots + A[j].$

A single element counts as a subarray (the sum is the value of that element). No elements counts as a subarray (the sum is 0).

1. Read and Understand the Problem



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• What is your **return type**? (Integer? List?)

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"subarray" means contiguous elements of the array

Key Idea with Divide and Conquer (and other recursive algorithms)

- If you identify that you want to use a recursive algorithm paradigm like Divide and Conquer, it's not enough to just answer those key questions on the previous slide
- Since you know you will have recursive calls, you need to be explicit about what those recursive calls are giving you that, when combined together, gives you the solution you're looking for
- You should be able to state a **clear English definition** of the return value you want to get from the recursive calls, keeping in mind the return type, the optimality, and the range & other parameters.

What is a **clear English definition** of the return value from the recursive calls?

Work through this problem with the people around you, and then we'll go over it together!

What is a **clear English definition** of the return value from the recursive calls?

maxSubarraySum(A[1..n]) returns the maximum sum of a
subarray, over all contiguous subarrays of A

2. Generate Examples



Good examples help with understanding now and testing later!

- You should generate two or three sample instances and the correct associated outputs.
- It's a good idea to have some "abnormal" examples consecutive negative numbers, very large negative numbers, only positive numbers, etc.
- *Note*: In general, you should not think of these examples as debugging examples null or the empty list is not a good example for this step. You can worry about edge cases at the end, once you have the main algorithm idea. You should be focused on the "typical" (not edge) case.

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[-3, -7, -2, -10] has a maximum subarray of [] with a sum of 0

[2, -100, 50, 3, -10, 17] has a maximum subarray of [50, 3, -10, 17] with a sum of 60

[1, 2, 3, 4] has a maximum subarray of [1, 2, 3, 4] with a sum of 10

[16, 20, -10, 4, 1,0] has a maximum subarray of [16, 20] with a sum of 36

[-1, 2, -3, -4] has a maximum subarray of [2] with a sum of 2

3. Come Up with a Baseline



Inefficient (non Divide and Conquer) First Attempt

- Review: In a time-constrained setting (like a **technical interview** or an **exam**) you often want a "baseline" algorithm. This should be an algorithm that you can implement and will give you the right answer, **even if it might be slow**.
- When you're pretty sure you want to use a Divide and Conquer algorithm, this step is **extremely** important! You need a (brute force) non Divide and Conquer baseline (with a quick runtime analysis) so you can see whether all the recursive steps of your Divide and Conquer algo are actually saving you any time!

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Runtime: There are $\mathcal{O}(n^2)$ different subarrays (pick start and end), and sum takes up to $\mathcal{O}(n)$ time per subarray, for a total of $\mathcal{O}(n^3)$.

```
function NaiveBaseline(A[1...n])
   bestSum ← -∞
   for i from 1 to n do
       for j from i to n do // j represents end index
            sum \leftarrow 0
                sum += A[k]
            if sum > bestSum then
                bestSum ← sum
   if bestSum < 0 then</pre>
       return 0
   return bestSum
```

```
// i represents start index
```

- **for** k from i to j **do** // Find sum for A[i]+...+A[j]

// handle all negative entries case

// empty subarray must be best here

```
function BetterBaseline(A[1..n])
bestSum ← -∞
for i from 1 to n do
    sum ← 0
    for j from i to n do
        sum += A[j]
        if sum > bestSum then
        bestSum ← sum
if bestSum < 0 then
        return 0
return bestSum</pre>
```

"Small" improvement: Use already computed sum from i to j to compute sum from i to j+1 This combines the sum procedure and the inner loop, so get an $\mathcal{O}(n^2)$ algorithm.

4. Brainstorm and Analyze Possible Algorithms



Think about How to Divide and Conquer

- Questions to help you brainstorm out your Divide and Conquer algo:
 - How do you want to split up the problem?
 - What is returned from the recursive calls? (hint: look back at part 1)
 - Imagine you have the answers from those recursive calls; what is there still to handle?
- When you have time, it's a good idea to try to run through your idea with some of the examples you came up with earlier, and see whether you get the correct output (especially as you try to transition from your brainstorming to formalizing your algorithm)

For each call SubarraySumDC(A[1..n]) answer these questions: How do you want to split up the problem?

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Let's just split the array in half! Make two recursive calls, one for each half (we don't know where the subarray is, so we'll have to make both).

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We get the maximum sum over all contiguous subarrays of A[1..n/2], and the maximum sum over all contiguous subarrays of A[n/2+1..n]

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It's possible that the subarray with the maximum sum actually "crosses" the two halves (i.e., includes both n/2 and n/2+1). We still need to discover and check this possibility.



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5. Write an Algorithm



Translate the brainstorm into an algorithm!

- We need to take those ideas we were just noodling on and write them into an algorithm!
- We can start with formalizing our ideas from earlier, but then we still need to figure out how to deal with those subarrays that cross from one half to the other...

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Key idea:

If we know that n/2 and n/2 + 1 are both included in the maximum subarray A[i..j], then we know that $i \le n/2$ and $j \ge n/2 + 1$.

So, *i* and *j* are "independent" of each other, and we can optimize for them separately. Now, we can have two separate loops instead of nested loops!


Loop upwards from middle

```
function bestSumFromMiddle(A[1..n])
best < -inf
partialSum < 0
for i < n/2+1 to n do
partialSum < partialSum + A[i]
if partialSum > output then
best < partialSum
return max(0, best)</pre>
```

6. Show Your Algorithm is Correct



Write a proof of correctness.

Hint: recursion \approx induction

IH is always strong induction style "my program is correct for all inputs of size $\leq k$ "

Work on this proof with the people around you, and then we'll go over it together!

We show that maxSubarraySum(A[1..n]) returns the max subarray sum by induction on n.

Base Case:

IH:

IS:

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Base Case: If n < 10, we run a brute force algorithm that checks every interval and returns the largest. Since every interval is checked, we return the largest.

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IS: Consider an array of length k + 1. We divide into cases, based on where the maximum subarray is.

<u>Case 1</u>: The maximum subarray is entirely in the left or right subarray.

Case 2: The maximum subarray crosses from the left to the right.

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<u>Case 1</u>: The maximum subarray is entirely in the left or right subarray.

By IH, the recursive calls will return the sum of largest subarray in each half. Furthermore, note that bestCrossSum is indeed the sum of some subarray. So, by the case assumption, bestCrossSum is not larger than the two recursive returns. Therefore, when we return the max of all three, we return the sum from the left or right subarray.

<u>Case 2</u>: The maximum subarray crosses from the left to the right.

Let the max subarray be A[i..j] which must have $i \le n/2$ and $j \ge n/2 + 1$ by the case assumption. Let the subarray picked by bestCrossSum be A[i'..j'] which is combined from bestSumToMiddle and bestSumFromMiddle.

bestSumToMiddle computes i' such that $A[i'] + \dots + A[n/2]$ is largest so $A[i'] + \dots + A[n/2] \ge A[i] + \dots + A[n/2]$ bestSumFromMiddle computes j' such that $A[n/2] + \dots + A[j']$ is largest so $A[n/2 + 1] + \dots + A[j'] \ge A[n/2 + 1] + \dots + A[j].$

Adding these two equations, we get $A[i'] + \dots + A[j'] \ge A[i] + \dots + A[j]$. Since A[i.j] is the max subarray, these two values must be equal. Hence, $A[i'] + \dots + A[j'] = A[i] + \dots + A[j]$ and we output bestCrossSum = $A[i'] + \dots + A[j'] = A[i] + \dots + A[j]$ as desired.

7. Optimize and Analyze the Run Time



Write the big-O of your code and justify the running time with a few sentences.

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Note that bestLeftSum and bestRightSum each require O(n) time to compute.

Since we make recursive calls on each half, we have the recurrence:

$$T(n) = \begin{cases} \mathcal{O}(1) & \text{if } n < 10\\ 2T(n/2) + \mathcal{O}(n) & \text{otherwise} \end{cases}$$

We have seen in class that this recurrence has the closed form $O(n \log n)$ (same as mergesort)

That's All, Folks!

Thanks for coming to section this week! Any questions?