

CS4102 Algorithms

Spring 2019

Warm up

Show that finding the minimum of an unordered list requires $\Omega(n)$ comparisons

Find Min, Lower Bound Proof

Show that finding the minimum of an unordered list requires $\Omega(n)$ comparisons

Suppose (toward contradiction) that there is an algorithm for Find Min that does fewer than $\frac{n}{2} = \Omega(n)$ comparisons.

This means there is at least one “uncompared” element
We can't know that this element wasn't the min!

2	8	19	20		3	9	-4
0	1	2	3	4	5	6	7

Announcements

- HW4 due Monday 3/4 at 11pm
 - Sorting
 - Written (use LaTeX!)
- No Instructor Office Hours this week
 - I'll be at SIGCSE
 - Available on Piazza and Email!
- HW1 solutions in-class on Wednesday
- Midterm next Wednesday
 - Covers material through today
 - Review session M or Tu evening

Today's Keywords

- Sorting
- Linear time Sorting
- Counting Sort
- Radix Sort
- Maximum Sum Continuous Subarray

CLRS Readings

- Chapter 8

Sorting in Linear Time

- Cannot be comparison-based
- Need to make some sort of assumption about the contents of the list
 - Small number of unique values
 - Small range of values
 - Etc.

Counting Sort

- **Idea:** Count how many things are less than each element

$$L = \begin{array}{|c|c|c|c|c|c|c|c|} \hline 3 & 6 & 6 & 1 & 3 & 4 & 1 & 6 \\ \hline 1 & 2 & 3 & 4 & 5 & 6 & 7 & 8 \\ \hline \end{array}$$

1. Range is $[1, k]$ (here $[1, 6]$)
make an array C of size k
populate with counts of each value

For i in L :
 $++C[L[i]]$

$$C = \begin{array}{|c|c|c|c|c|c|} \hline 2 & 0 & 2 & 1 & 0 & 3 \\ \hline 1 & 2 & 3 & 4 & 5 & 6 \\ \hline \end{array}$$

running sum



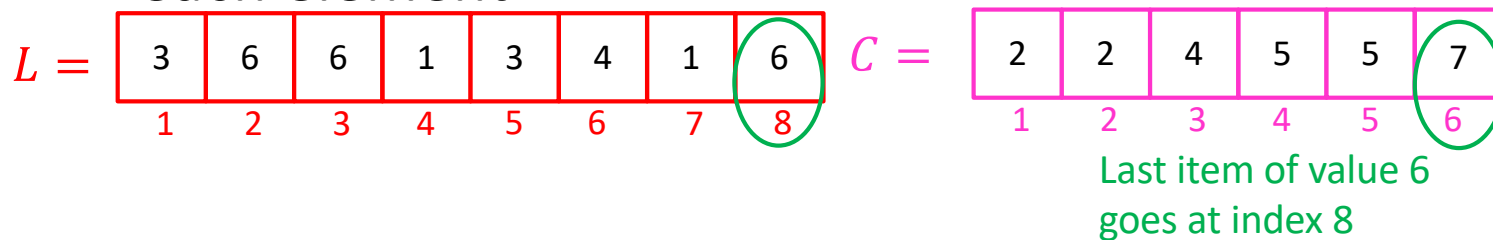
2. Take “running sum” of C
to count things less than each value

$$C = \begin{array}{|c|c|c|c|c|c|} \hline 2 & 2 & 4 & 5 & 5 & 8 \\ \hline 1 & 2 & 3 & 4 & 5 & 6 \\ \hline \end{array}$$

To sort: last item of
value 3 goes at index 4

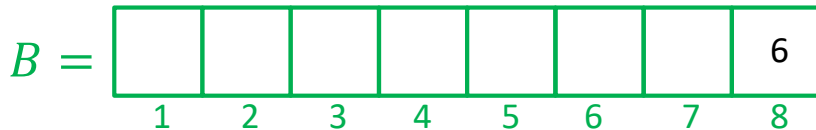
Counting Sort

- Idea: **Count** how many things are less than each element



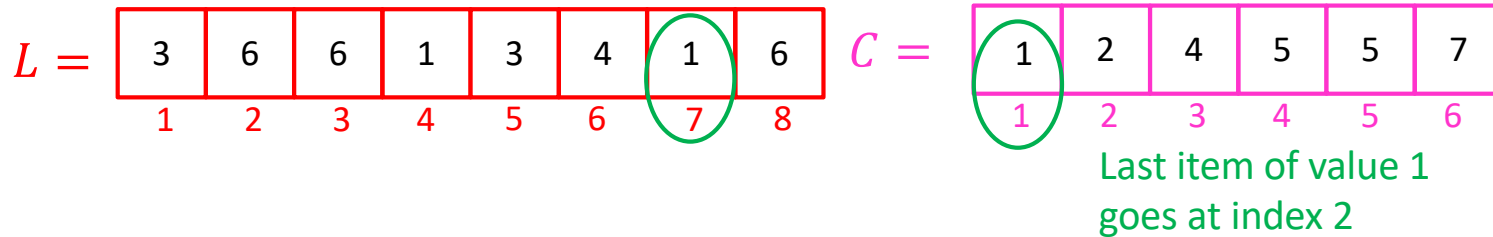
For each **element** of L (last to first):
 Use C to find its **proper place** in B
 Decrement that position of C

For $i = \text{len}(L)$ downto 1:
 $B[C[L[i]]] = L[i]$
 $C[L[i]] = C[L[i]] - 1$



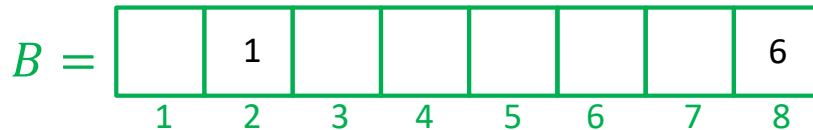
Counting Sort

- Idea: **Count** how many things are less than each element



For each **element** of L (last to first):
 Use C to find its **proper place** in B
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For $i = \text{len}(L)$ downto 1:
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 $C[L[i]] = C[L[i]] - 1$



Run Time: $O(n + k)$

Memory: $O(n + k)$

Counting Sort

- Why not always use counting sort?
- For 64-bit numbers, requires an array of length $2^{64} > 10^{19}$
 - 5 GHz CPU will require > 116 years to initialize the array
 - 18 Exabytes of data
 - Total amount of data that Google has

12 Exabytes



Radix Sort

- Idea: **Stable sort** on each digit, from least significant to most significant

103	801	401	323	255	823	999	101	113	901	555	512	245	800	018	121
0	1	2	3	4	5	6	7	8	9	10	11	12	13	14	15

Place each element into a “bucket” according to its 1’s place

800	801 401 101 901 121	512	103 323 823 113		255 555 245			018	999
0	1	2	3	4	5	6	7	8	9

Radix Sort

- Idea: **Stable sort** on each digit, from least significant to most significant

Place each element into a “bucket” according to its 10’s place

800	801 401 101 901 121	512	103 323 823 113		255 555 245			018	999
0	1	2	3	4	5	6	7	8	9

800									
801	512	121							
401	113	323		245	255				999
101	018	823			555				
901									
103									
0	1	2	3	4	5	6	7	8	9

Radix Sort

- Idea: **Stable sort** on each digit, from least significant to most significant

Place each element into a "bucket" according to its 100's place

800										
801	512	121								
401	113	323		245	255				999	
101					555					
901	018	823								
103										
	0	1	2	3	4	5	6	7	8	9

Run Time: $O(d(n + b))$
 d = digits in largest value
 b = base of representation

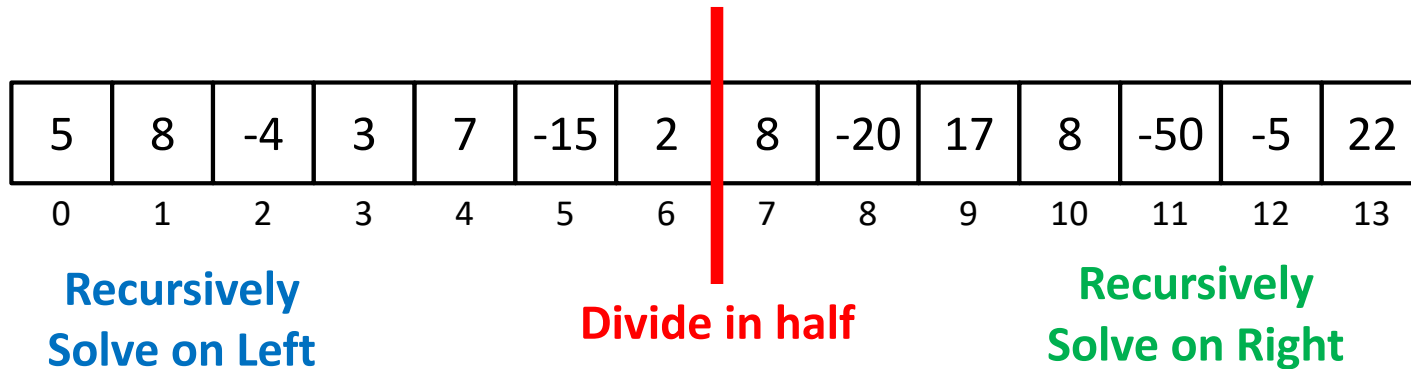
	101							800	901	
018	103	245	323	401	512		801	823	999	
	113	255			555					
	121									
	0	1	2	3	4	5	6	7	8	9

Maximum Sum Continuous Subarray Problem

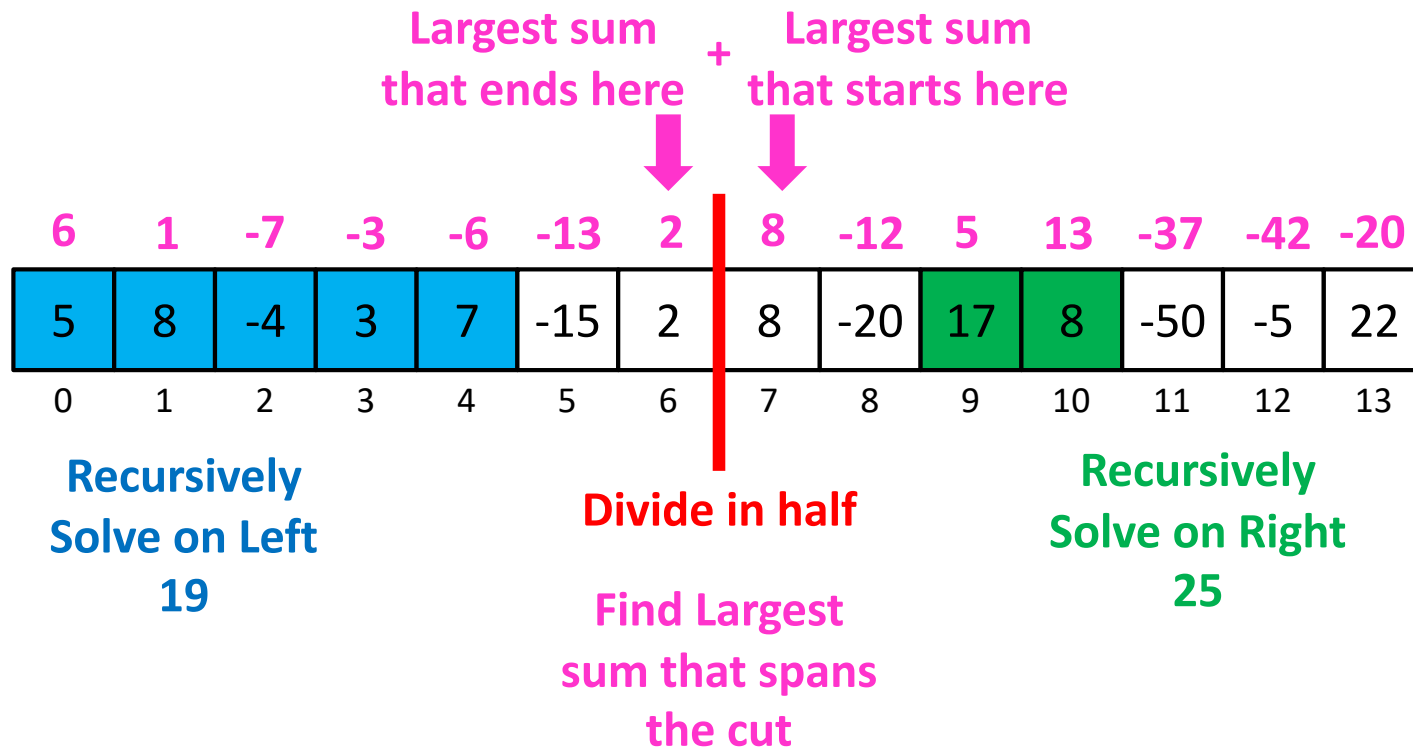
The maximum-sum subarray of a given array of integers A is the interval $[a, b]$ such that the sum of all values in the array between a and b inclusive is maximal.

Given an array of n integers (may include both positive and negative values), give a $O(n \log n)$ algorithm for finding the maximum-sum subarray.

Divide and Conquer $\Theta(n \log n)$

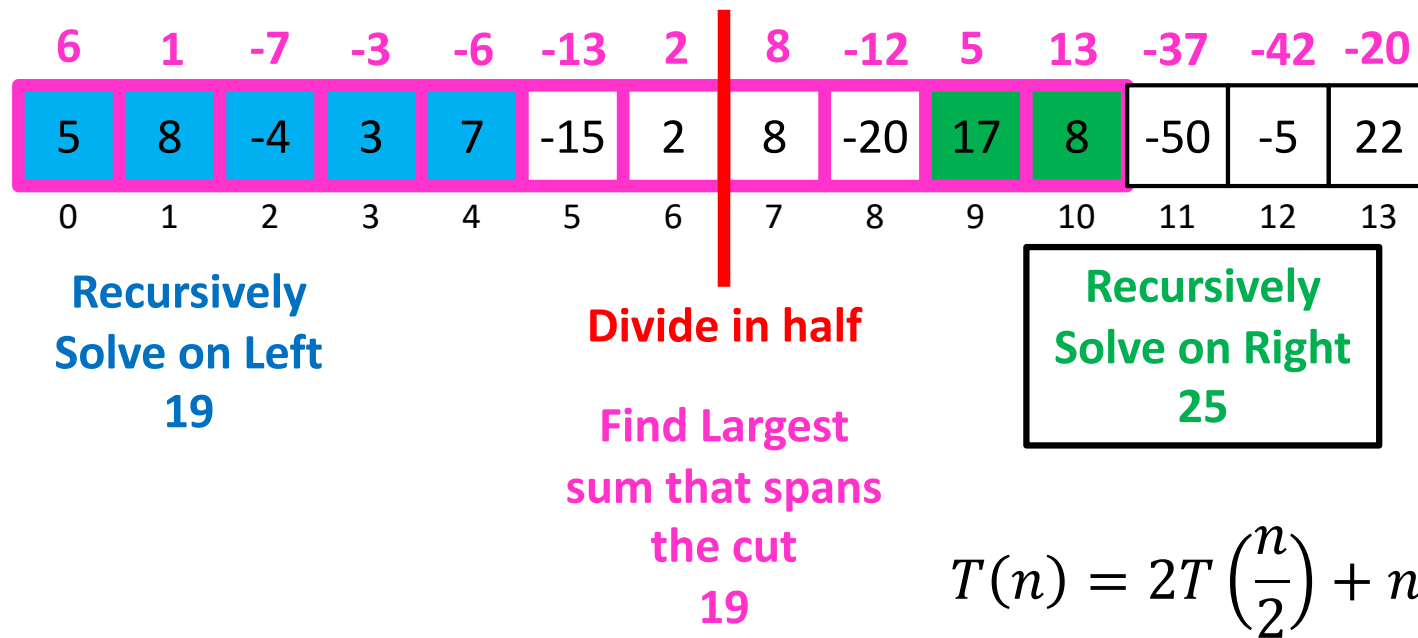


Divide and Conquer $\Theta(n \log n)$



Divide and Conquer $\Theta(n \log n)$

Return the Max of
Left, Right, Center



Divide and Conquer Summary

Typically multiple subproblems.

Typically all roughly the same size.

- **Divide**
 - Break the list in half
- **Conquer**
 - Find the best subarrays on the left and right
- **Combine**
 - Find the best subarray that “spans the divide”
 - I.e. the best subarray that ends at the divide concatenated with the best that starts at the divide

Generic Divide and Conquer Solution

```
def myDCalgo(problem):  
    if baseCase(problem):  
        solution = solve(problem) #brute force if necessary  
        return solution  
    subproblems = Divide(problem)  
    for sub in subproblems:  
        subsolutions.append(myDCalgo(sub))  
    solution = Combine(subsolutions)  
    return solution
```

MSCS Divide and Conquer $\Theta(n \log n)$

```
def MSCS(list):  
    if list.length < 2:  
        return list[0]    #list of size 1 the sum is maximal  
    {listL, listR} = Divide (list)  
    for list in {listL, listR}:  
        subSolutions.append(MSCS(list))  
    solution = max(solnL, solnR, span(listL, listR))  
    return solution
```

Types of “Divide and Conquer”

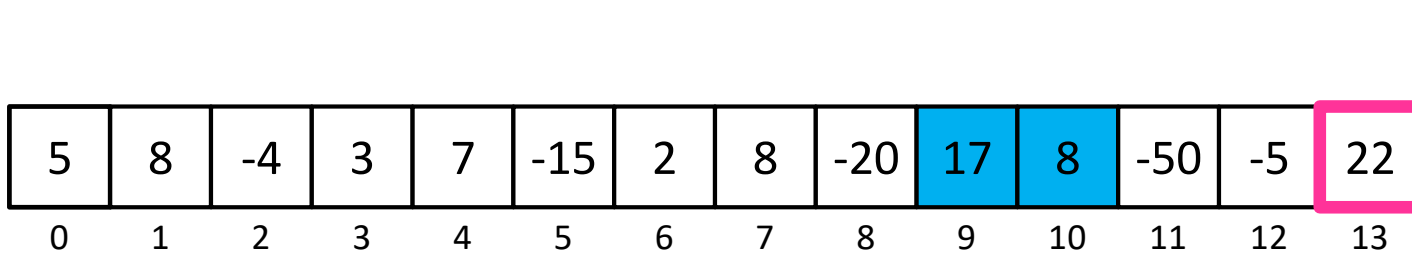
- Divide and Conquer
 - Break the problem up into several subproblems of roughly equal size, recursively solve
 - E.g. Karatsuba, Closest Pair of Points, Mergesort...
- Decrease and Conquer
 - Break the problem into a single smaller subproblem, recursively solve
 - E.g. Gotham City Police, Quickselect, Binary Search

Pattern So Far

- Typically looking to divide the problem by some fraction ($\frac{1}{2}$, $\frac{1}{4}$ the size)
- Not necessarily always the best!
 - Sometimes, we can write faster algorithms by finding **unbalanced** divides.

Unbalanced Divide and Conquer

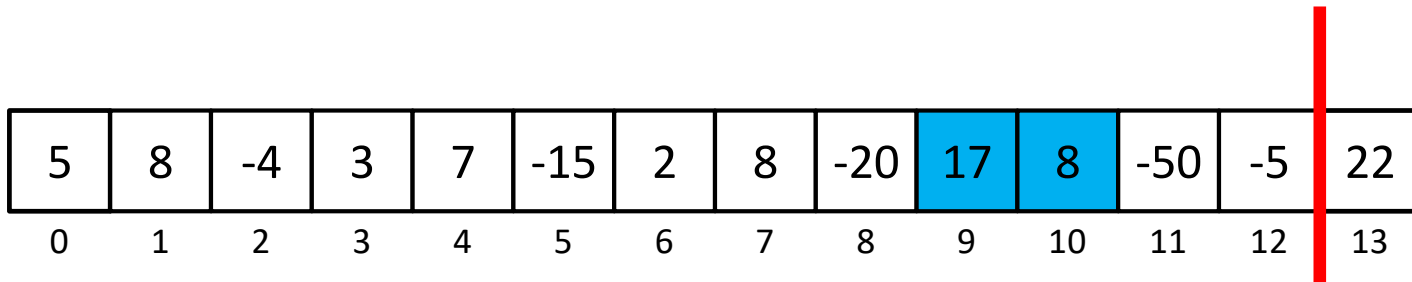
- **Divide**
 - Make a subproblem of all but the last element
- **Conquer**
 - Find best subarray on the left ($BSL(n - 1)$)
 - Find the best subarray ending at the divide ($BED(n - 1)$)
- **Combine**
 - New Best Ending at the Divide:
 - $BED(n) = \max(BED(n - 1) + arr[n], 0)$
 - New best on the left:
 - $BSL(n) = \max(BSL(n - 1), BED(n))$



**Recursively
Solve on Left
25**

**Find Largest
sum ending at
the cut
22**

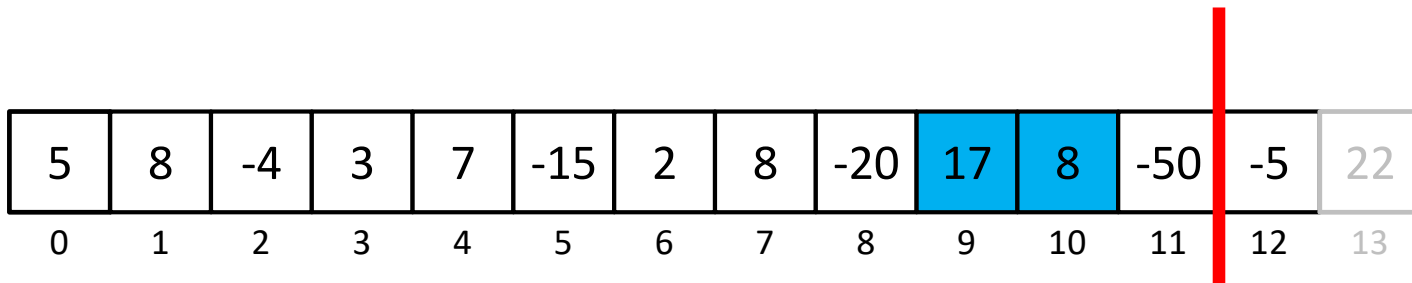
Divide



**Recursively
Solve on Left
25**

**Find Largest
sum ending at
the cut
0**

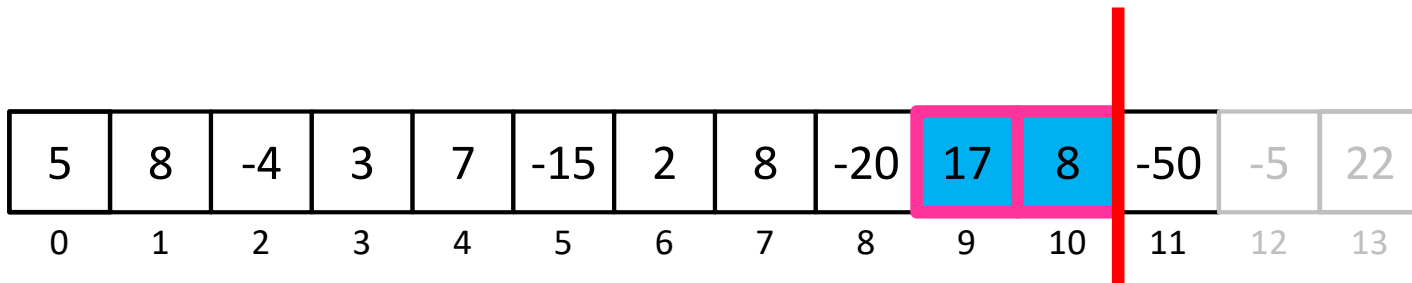
Divide



**Recursively
Solve on Left
25**

**Find Largest
sum ending at
the cut
0**

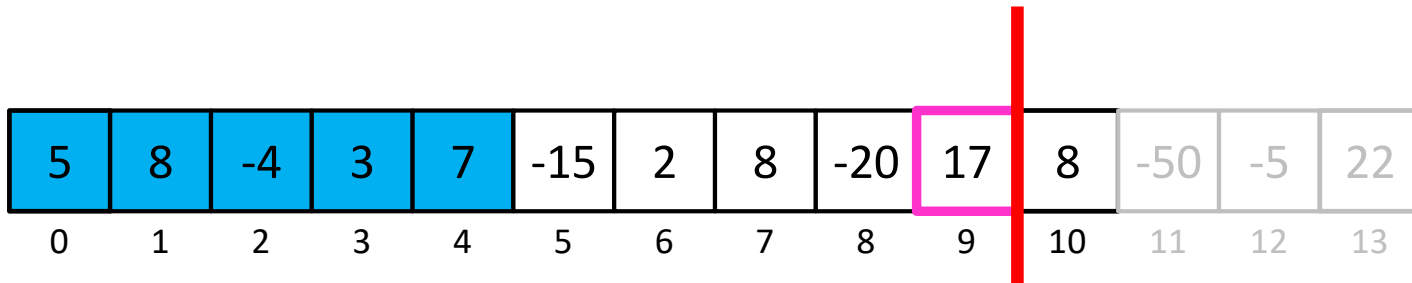
Divide



**Recursively
Solve on Left
25**

Divide

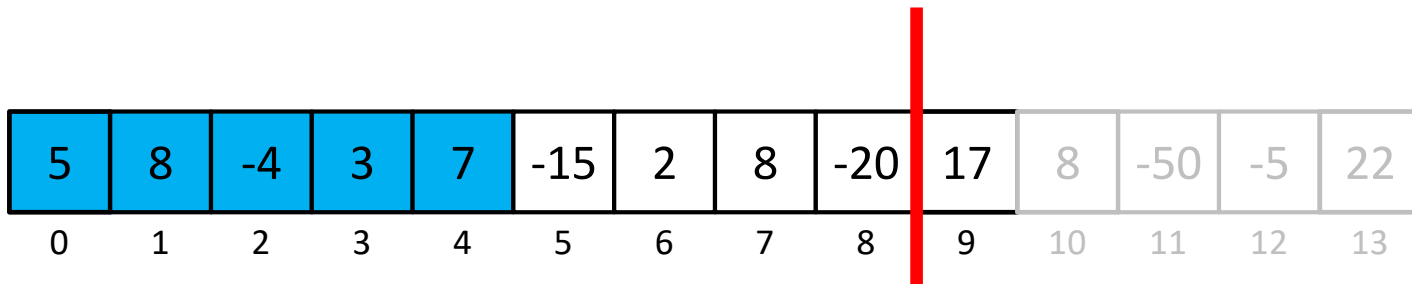
**Find Largest
sum ending at
the cut
25**



**Recursively
Solve on Left
19**

Divide

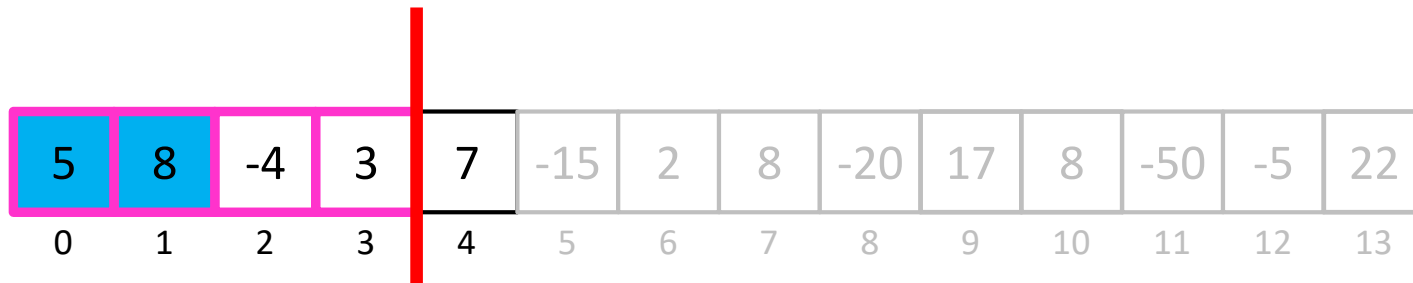
**Find Largest
sum ending at
the cut
17**



**Recursively
Solve on Left
19**

Divide

**Find Largest
sum ending at
the cut
0**



Recursively
Solve on Left
13

Divide

Find Largest
sum ending at
the cut
12

Unbalanced Divide and Conquer

- **Divide**
 - Make a subproblem of all but the last element
- **Conquer**
 - Find best subarray on the left ($BSL(n - 1)$)
 - Find the best subarray ending at the divide ($BED(n - 1)$)
- **Combine**
 - New Best Ending at the Divide:
 - $BED(n) = \max(BED(n - 1) + arr[n], 0)$
 - New best on the left:
 - $BSL(n) = \max(BSL(n - 1), BED(n))$

Was unbalanced better? YES

- Old:
 - We divided in **Half**
 - We solved 2 different problems:
 - Find the best overall on **BOTH** the **left/right**
 - Find the best which end/start on **BOTH** the **left/right** respectively
 - **Linear** time combine
- New:
 - We divide by **1, n-1**
 - We solve 2 different problems:
 - Find the best overall on the **left ONLY**
 - Find the best which ends on the **left ONLY**
 - **Constant** time combine

$$T(n) = 2T\left(\frac{n}{2}\right) + n$$

$$T(n) = \Theta(n \log n)$$

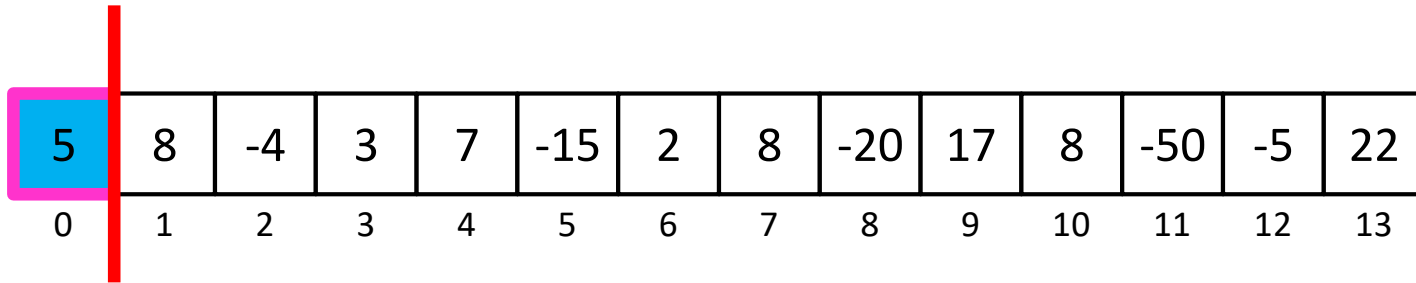
$$T(n) = 1T(n-1) + 1$$

$$T(n) = \Theta(n)$$

Maximum Sum Continuous Subarray Problem Redux

- Solve in $O(n)$ by increasing the problem size by 1 each time.
- **Idea:** Only include negative values if the positives on both sides of it are “worth it”

$\Theta(n)$ Solution



Begin here

Remember two values:

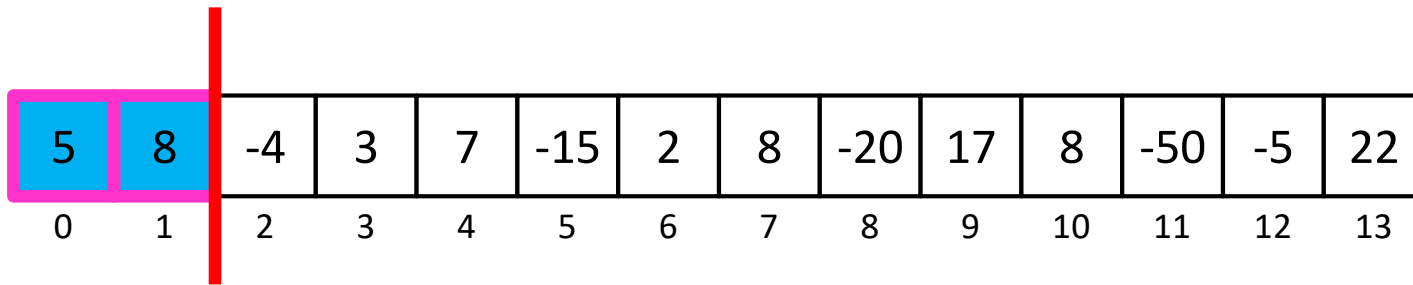
Best So Far

5

Best ending here

5

$\Theta(n)$ Solution

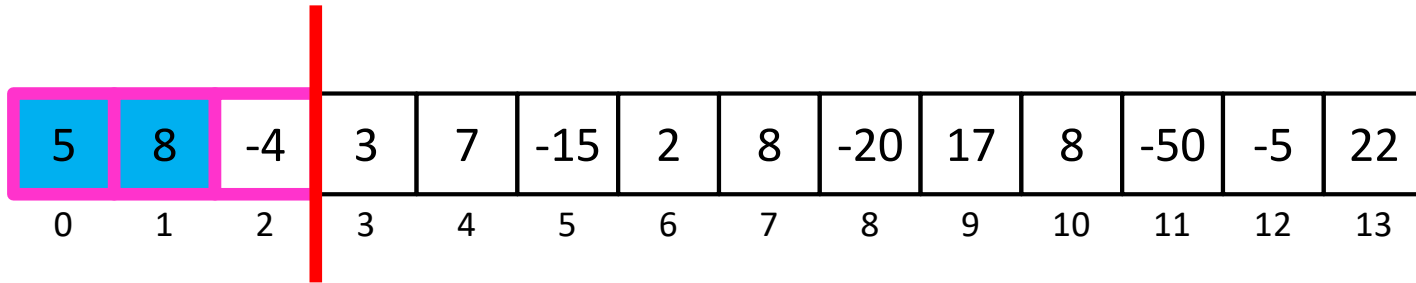


Remember two values:

Best So Far
13

Best ending here
13

$\Theta(n)$ Solution

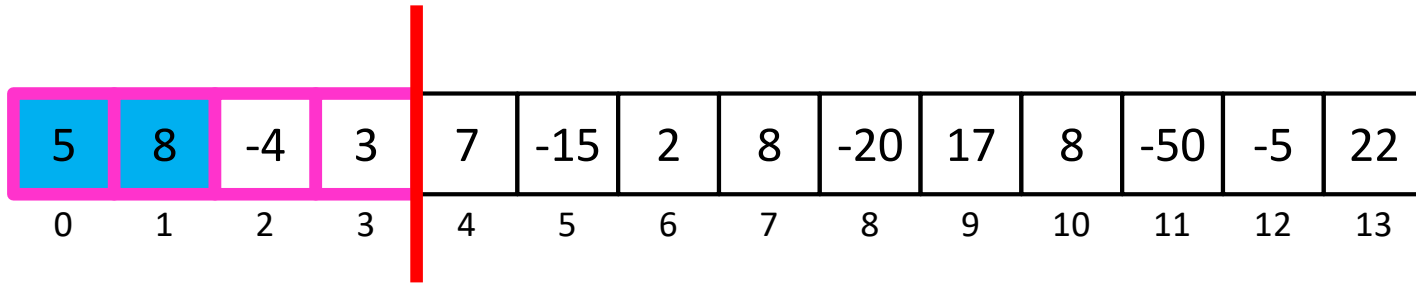


Remember two values:

Best So Far
13

Best ending here
9

$\Theta(n)$ Solution

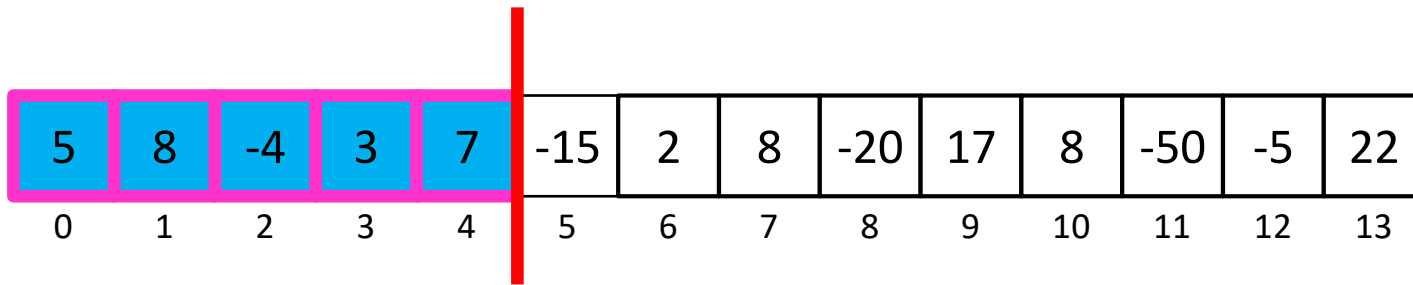


Remember two values:

Best So Far
13

Best ending here
12

$\Theta(n)$ Solution

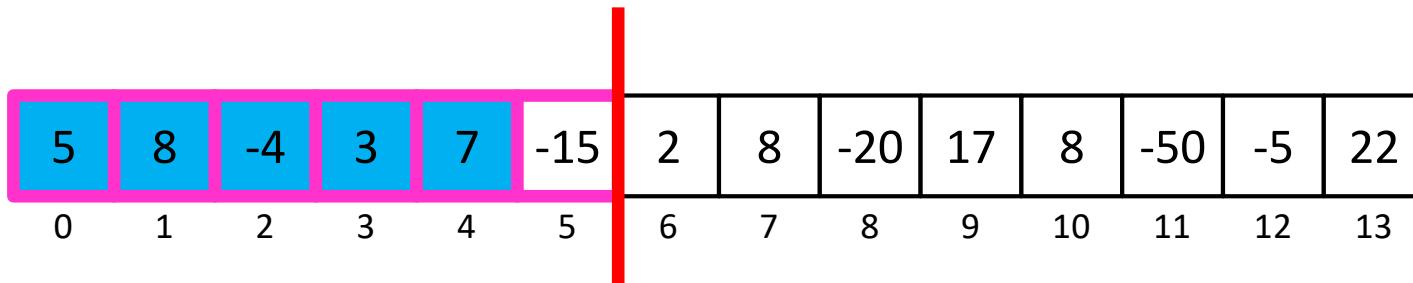


Remember two values:

Best So Far
19

Best ending here
19

$\Theta(n)$ Solution

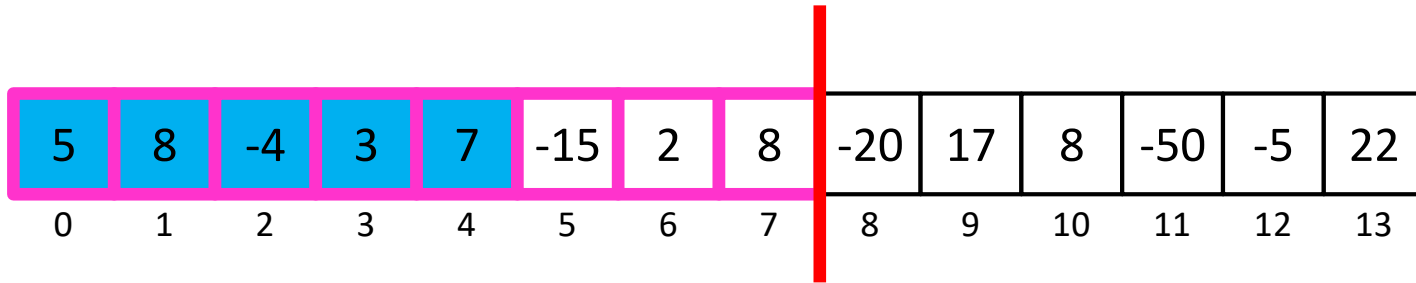


Remember two values:

Best So Far
19

Best ending here
4

$\Theta(n)$ Solution

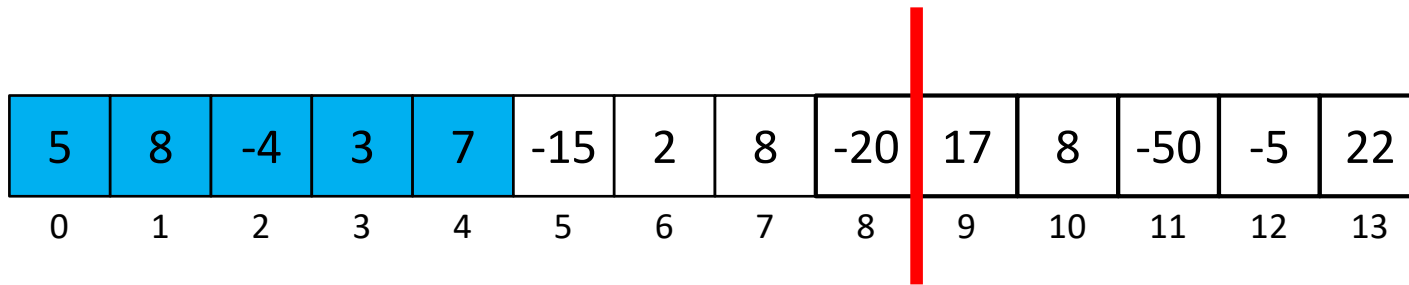


Remember two values:

Best So Far
19

Best ending here
14

$\Theta(n)$ Solution

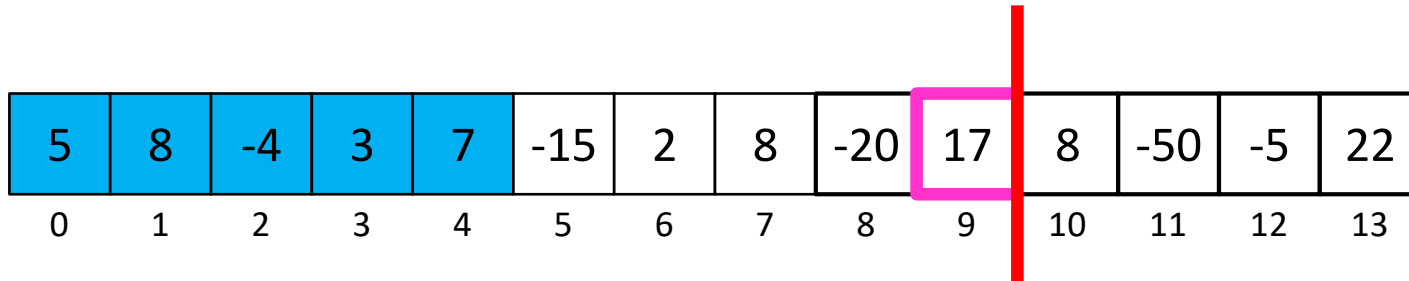


Remember two values:

Best So Far
19

Best ending here
0

$\Theta(n)$ Solution

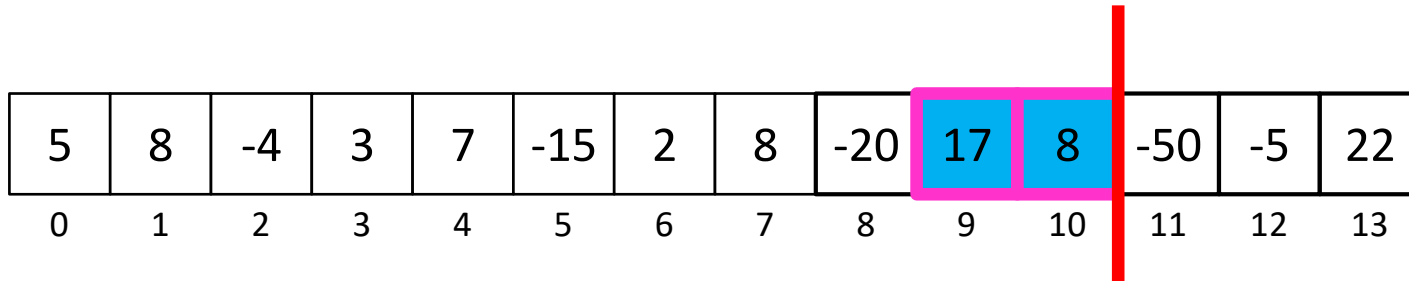


Remember two values:

Best So Far
19

Best ending here
17

$\Theta(n)$ Solution



Remember two values:

Best So Far
25

Best ending here
25

End of Midterm Exam Materials!

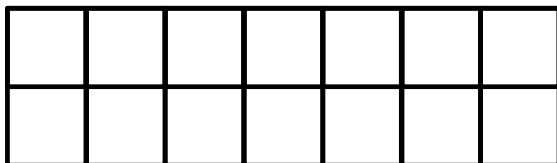


"Mr. Osborne, may I be excused? My brain is full."

Mid-Class Stretch

How many ways are there to tile a $2 \times n$ board with dominoes?

How many ways to
tile this:



With these?

