## CS4102 Algorithms

Spring 2020 – Horton's Slides

#### Warm up

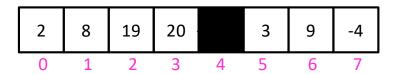
Show that finding the minimum of an unordered list requires  $\Omega(n)$  comparisons

### 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!



#### Homeworks

- HW4 due 11pm Thursday, February 27, 2020
  - Divide and Conquer and Sorting
  - Written (use LaTeX!)
  - Submit BOTH a pdf and a zip file (2 separate attachments)
- Midterm: March 4 (two weeks away!)
- Regrade Office Hours
  - Fridays 2:30pm-3:30pm (Rice 210)

## Today's Keywords

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

## CLRS Readings

• Chapter 8

## Sorting, so far

Sorting algorithms we have discussed:

```
- Mergesort O(n \log n) Optimal!
```

- Quicksort  $O(n \log n)$  Optimal!

Other sorting algorithms (will discuss):

```
- Bubblesort O(n^2)
```

- Insertionsort  $O(n^2)$ 

- Heapsort  $O(n \log n)$  Optimal!

## Speed Isn't Everything

#### Important properties of sorting algorithms:

- Run Time
  - Asymptotic Complexity
  - Constants
- In Place (or In-Situ)
  - Done with only constant additional space
- Adaptive
  - Faster if list is nearly sorted
- Stable
  - Equal elements remain in original order
- Parallelizable
  - Runs faster with multiple computers

## Mergesort

- Divide:
  - Break *n*-element list into two lists of n/2 elements
- Conquer:
  - If n > 1: Sort each sublist recursively
  - If n = 1: List is already sorted (base case)
- Combine:
  - Merge together sorted sublists into one sorted list

In Place?Adaptive?Stable?NoNoYes!(usually)

Run Time?  $\Theta(n \log n)$  Optimal!

## Merge

- Combine: Merge sorted sublists into one sorted list
- We have:
  - 2 sorted lists ( $L_1$ ,  $L_2$ )
  - -1 output list ( $L_{out}$ )

```
While (L_1 and L_2 not empty):
```

```
If L_1[0] \le L_2[0]:
L_{out}.\mathsf{append}(L_1.\mathsf{pop}())
Else:
\mathsf{comes first}
```

 $L_{out}$ .append( $L_2$ .pop())

 $L_{out}$ .append( $L_1$ )

 $L_{out}$ .append( $L_2$ )

## Mergesort

- Divide:
  - Break *n*-element list into two lists of n/2 elements
- Conquer:
  - If n > 1: Sort each sublist recursively
  - If n = 1: List is already sorted (base case)
- Combine:
  - Merge together sorted sublists into one sorted list

Run Time?  $\Theta(n \log n)$  Optimal!

In Place? Adaptive? Stable?

No No Yes!

(usually)

Parallelizable?
Yes!

## Mergesort

#### Divide:

- Break *n*-element list into two lists of n/2 elements

Parallelizable:
Allow different
machines to work
on each sublist

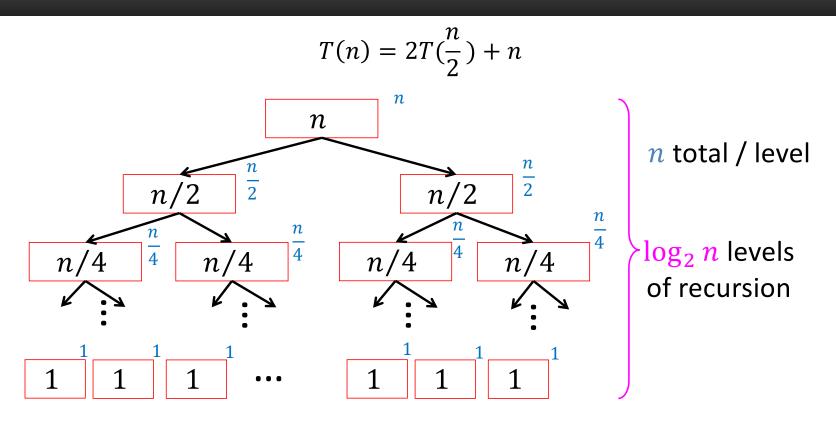
#### Conquer:

- If n > 1:
  - Sort each sublist recursively
- If n = 1:
  - List is already sorted (base case)

#### • Combine:

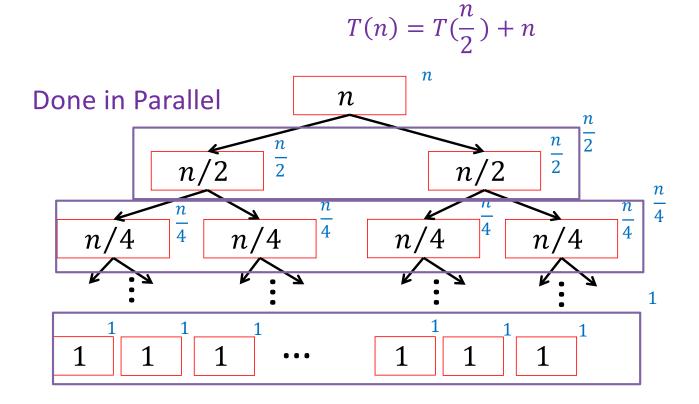
Merge together sorted sublists into one sorted list

## Mergesort (Sequential)



Run Time:  $\Theta(n \log n)$ 

## Mergesort (Parallel)



Run Time:  $\Theta(n)$ 

#### Quicksort

Idea: pick a partition element, recursively sort two sublists around that element

- Divide: select an element p, Partition(p)
- Conquer: recursively sort left and right sublists
- Combine: Nothing!

Run Time?

 $\Theta(n \log n)$ 

(almost always)

Better constants

than Mergesort

In Place?

Adaptive?

Stable?

Parallelizable?

kinda

No!

No

Yes!

Uses stack for recursive calls

### Bubble Sort

Idea: March through list, swapping adjacent elements if out of order, repeat until sorted

8	5	7	9	12	10	1	2	4	3	6	11
5	8	7	9	12	10	1	2	4	3	6	11
5		8	9	12	10	1	2	4	3	6	11
5	7	8	9	12	10	1	2	4	3	6	11

#### Bubble Sort

 Idea: March through list, swapping adjacent elements if out of order, repeat until sorted **Run Time?** 

 $\Theta(n^2)$ 

Constants worse than Insertion Sort

In Place?

Adaptive?

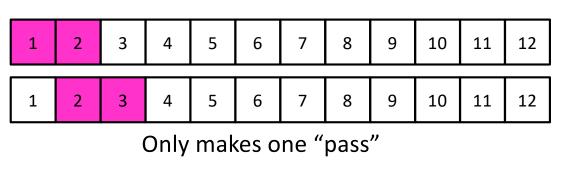
Yes

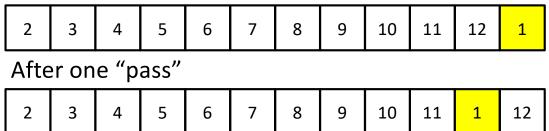
Kinda

"Compared to straight insertion [...], bubble sorting requires a more complicated program and takes about twice as long!" –Donald Knuth

## Bubble Sort is "almost" Adaptive

Idea: March through list, swapping adjacent elements if out of order





Requires n passes, thus is  $O(n^2)$ 

#### Bubble Sort

 Idea: March through list, swapping adjacent elements if out of order, repeat until sorted Run Time?  $\Theta(n^2)$ 

Constants worse

than Insertion Sort

Parallelizable?

No

In Place?

Adaptive?

Stable?

Yes!

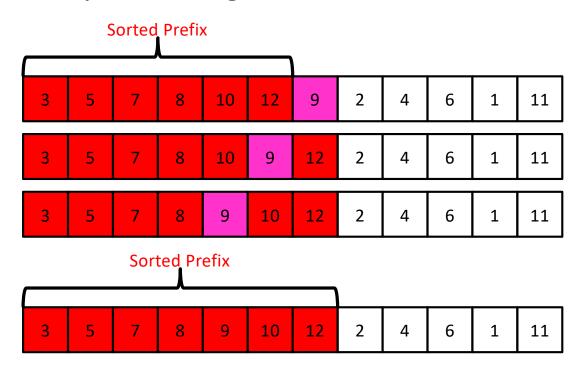
Kinda Not really Yes

"the bubble sort seems to have nothing to recommend it, except a catchy name and the fact that it leads to some interesting theoretical problems" –Donald Knuth, The Art of Computer Programming



#### Insertion Sort

Idea: Maintain a sorted list prefix, extend that prefix by "inserting" the next element



## Insertion Sort

 Idea: Maintain a sorted list prefix, extend that prefix by "inserting" the next element **Run Time?** 

 $\Theta(n^2)$ 

(but with very small constants)

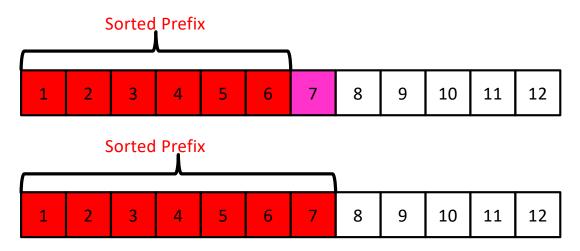
Great for short lists!

<u>In Place?</u> Adaptive?

Yes! Yes

## Insertion Sort is Adaptive

Idea: Maintain a sorted list prefix, extend that prefix by "inserting" the next element



Only one comparison needed per element! Runtime: O(n)

#### Insertion Sort

 Idea: Maintain a sorted list prefix, extend that prefix by "inserting" the next element Run Time?

 $\Theta(n^2)$ 

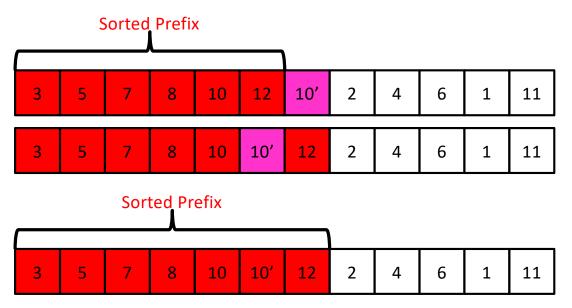
(but with very small constants)

Great for short lists!

<u>In Place?</u> <u>Adaptive?</u> <u>Stable?</u> Yes! Yes Yes

#### Insertion Sort is Stable

 Idea: Maintain a sorted list prefix, extend that prefix by "inserting" the next element



The "second" 10 will stay to the right

#### Insertion Sort

 Idea: Maintain a sorted list prefix, extend that prefix by "inserting" the next element Run Time?  $\Theta(n^2)$ 

(but with very small constants)
Great for short lists!

In Place?

Adaptive?

Stable?

Parallelizable?

Yes!

Yes

Yes

No

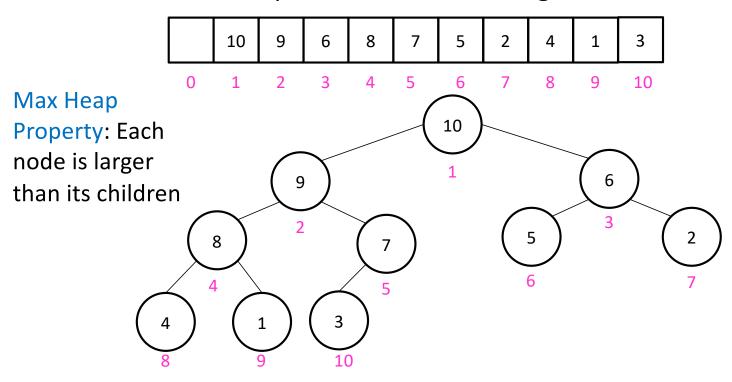
"All things considered, it's actually a pretty good sorting algorithm!" –Nate Brunelle

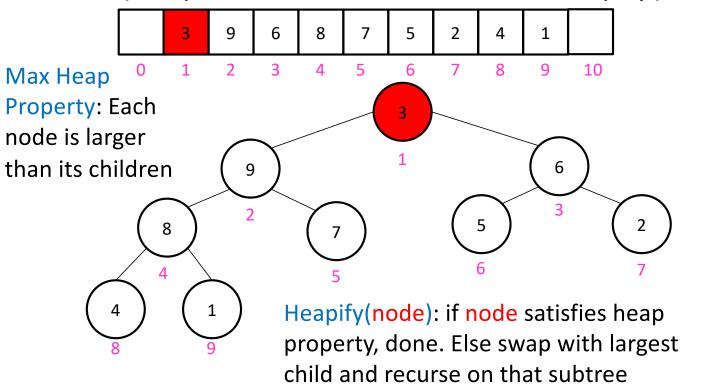
Can sort a list as it is received, i.e., don't need the entire list to begin sorting

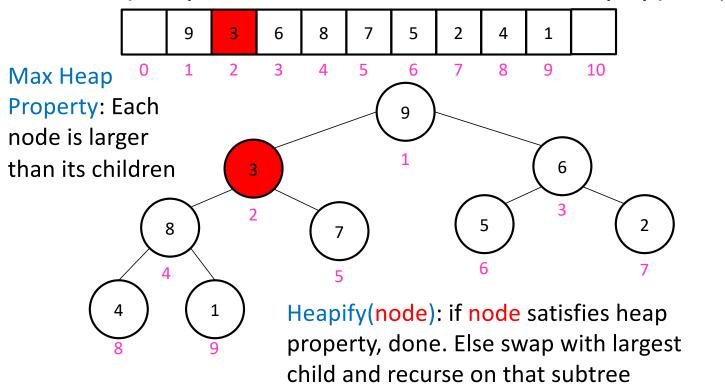
Online?

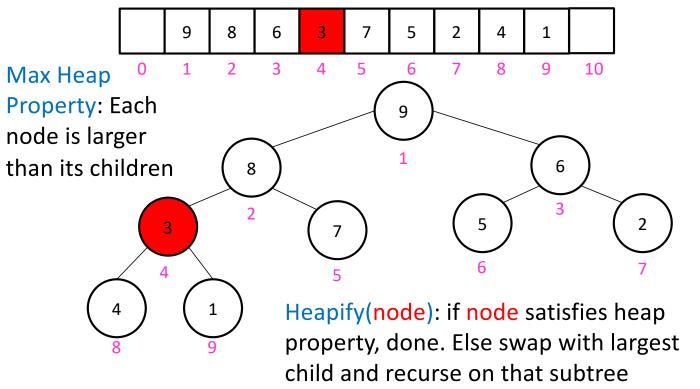
Yes

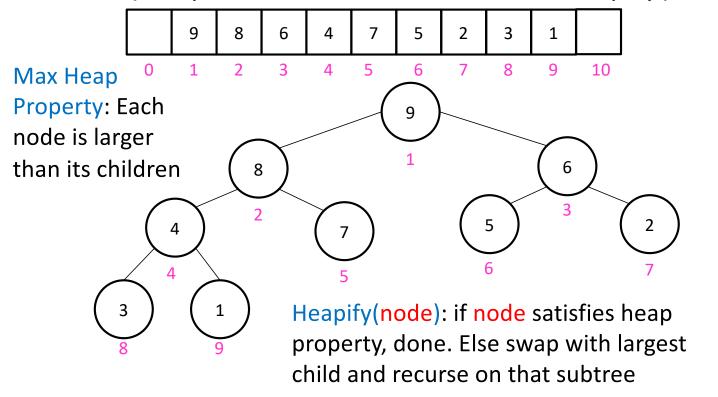
• Idea: Build a Heap, repeatedly extract max element from the heap to build sorted list Right-to-Left







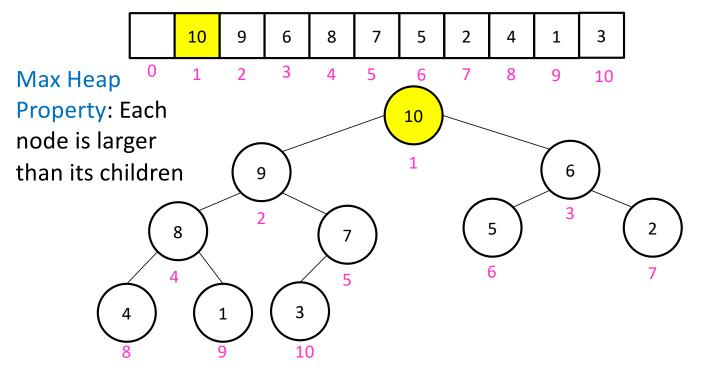


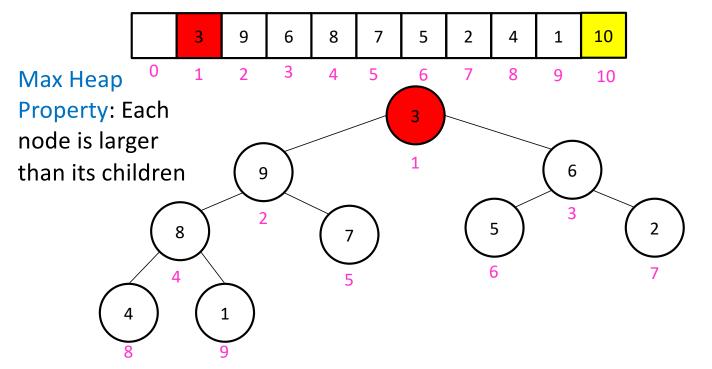


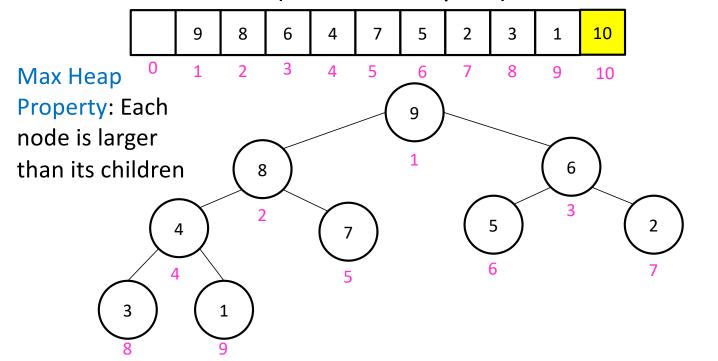
 Idea: Build a Heap, repeatedly extract max element from the heap to build sorted list Rightto-Left Run Time?  $\Theta(n \log n)$ Constants worse than Quick Sort

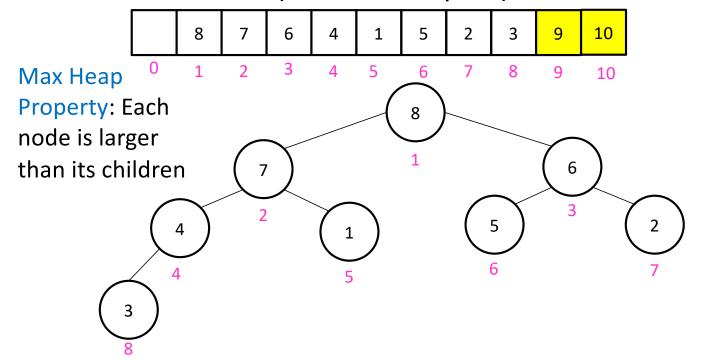
In Place?

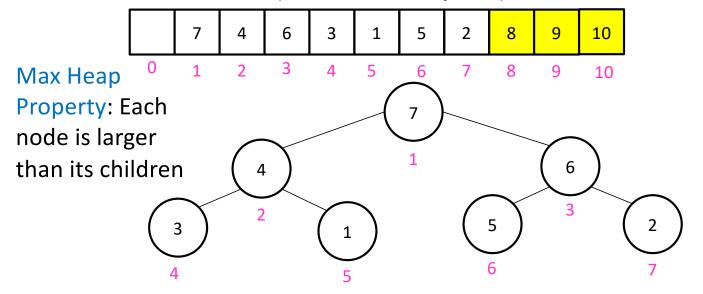
Yes!











 Idea: Build a Heap, repeatedly extract max element from the heap to build sorted list Rightto-Left Run Time?  $\Theta(n \log n)$ Constants worse
than Quick Sort
Parallelizable?

<u>In Place?</u> <u>Adaptive?</u> <u>Stable?</u> Yes! No No

No

## Sorting, so far

Sorting algorithms we have discussed:

```
- Mergesort O(n \log n) Optimal!
```

- Quicksort  $O(n \log n)$  Optimal!

Other sorting algorithms (will discuss):

```
- Bubblesort O(n^2)
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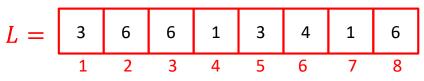
- Insertionsort  $O(n^2)$ 

- Heapsort  $O(n \log n)$  Optimal!

### Sorting in Linear Time

- Sometimes we can sort in linear time!
  - Wait, what? We used decision trees to prove sorting is  $\Omega(n \log n)$
  - Remember: proof assumed key-comparison was our basic operation
- Thus, if we can do something more than just compare two keys, then...
  - Similar situation: binary search is optimal, but hashing can be faster
- Possible approach: make some sort of assumption about the contents of the list
  - Small number of unique values
  - Small range of values
  - Etc.
- Cannot be comparison-based! We see examples that use a key's numeric value.

Starting point: Determine how often each element occurs



Note: if range is [0,k] then C will be zero-index.

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

$$C = \begin{bmatrix} 2 & 0 & 2 & 1 & 0 & 3 \\ 1 & 2 & 3 & 4 & 5 & 6 \end{bmatrix}$$

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

2.We could easily use *C* to produce a list of sorted key values in a list *B*. Do you see how? (Discuss.)

That's sorting, isn't it, so that's all we need, right? (Answer: no.)

First Idea: Use index and Count to create output list of key values

$$L = \begin{bmatrix} 3 & 6 & 6 & 1 & 3 & 4 & 1 & 6 \\ & 1 & 2 & 3 & 4 & 5 & 6 & 7 & 8 \end{bmatrix}$$

$$C = \begin{bmatrix} 2 & 0 & 2 & 1 & 0 & 3 \\ & 1 & 2 & 3 & 4 & 5 & 6 \end{bmatrix}$$

```
b_idx = 1  // next position in output list B
for i = 1 to len(C):  // look at count for next value in range
  for j = 1 to C[i]:  // for each time i occurs...
     B[b_idx] = i  // ...put value i into output list
     b_idx = b_idx + 1
```

Complexity:  $\Theta(\max(n, k)) = \Theta(n + k)$ 

What's wrong with this approach?

Data associated with keys? Stable?

Priority	Associated Data
2	Data, 1st with pri=2
1	Data, 1 <sup>st</sup> with pri=1
2	Data, 2 <sup>nd</sup> with pri=2
1	Data, 2 <sup>nd</sup> with pri=1

Better Idea: Count how many things are ≤ each element

$$L = \begin{bmatrix} 3 & 6 & 6 & 1 & 3 & 4 & 1 & 6 \\ & 1 & 2 & 3 & 4 & 5 & 6 & 7 & 8 \end{bmatrix}$$

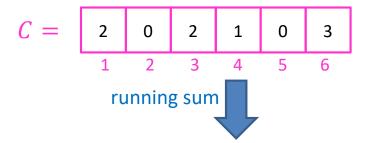
1.Range is [1, k] (here [1,6]) make an array C of size k populate with counts of each value

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

2.Take "running sum" of *C* to count things less than each value

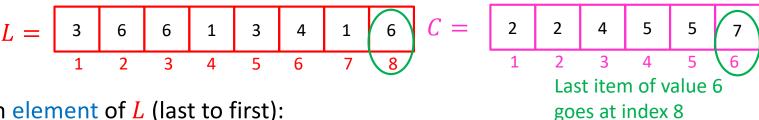
For 
$$i = 2$$
 to len( $C$ ):  

$$C[i] = C[i-1] + C[i]$$



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

Idea: Count how many things are ≤ each element



For each element of *L* (last to first):

Use *C* to find its proper place in *B*.

Put element from L there.

Decrement that position of C.

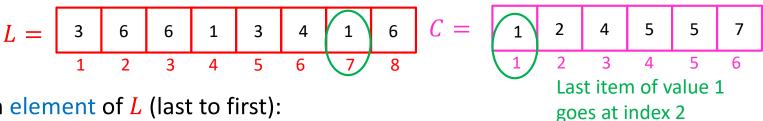
Why? If earlier element in L with same value is found, placed right before it.

For 
$$i = \text{len}(\underline{L})$$
 downto 1:  

$$B\left[C[L[i]]\right] = L[i]$$

$$C[L[i]] = C[L[i]] - 1$$

Idea: Count how many things are less than each element



For each element of *L* (last to first):

Use  $\mathcal{C}$  to find its proper place in  $\mathcal{B}$ .

Put element from L there.

Decrement that position of C.

Why? If earlier element in L with same value is found, placed right before it.

For 
$$i = \text{len}(\underline{L})$$
 downto 1:  

$$B \left[ C[\underline{L}[i]] \right] = \underline{L}[i]$$

$$C[\underline{L}[i]] = C[\underline{L}[i]] - 1$$

Run Time: O(n + k)Memory: O(n + k)

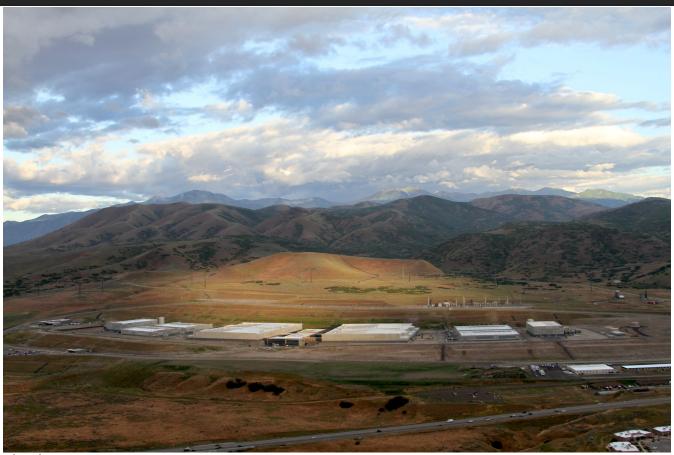
Is this stable? Why or why not?

- 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 (?)

One Exabyte = 10<sup>18</sup> bytes 1 million terabytes (TB) 1 billion gigabytes (GB)

100,000 x Library of Congress (print)

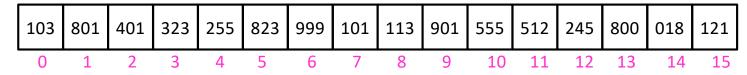
# 12 Exabytes



https://en.wikipedia.org/wiki/Utah\_Data\_Center

### Radix Sort

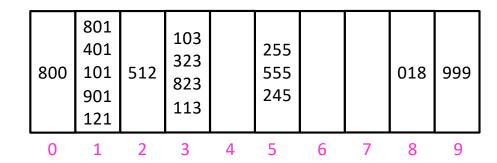
 Idea: Stable sort on each digit, from least significant to most significant



Place each element into a "bucket" according to its 1's place

Which stable sort would you choose?

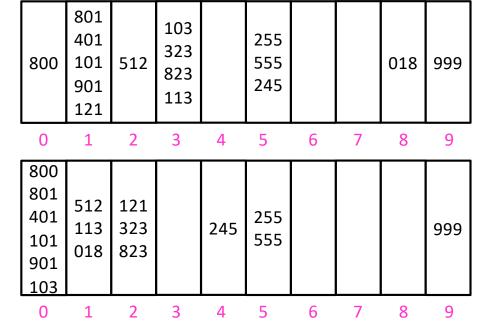
Read CLRS, Section 8.3!



### 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



### Radix Sort

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

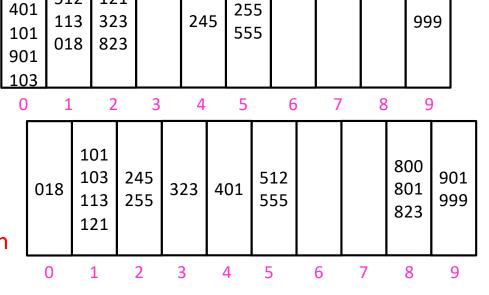
512

121

800 801

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

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

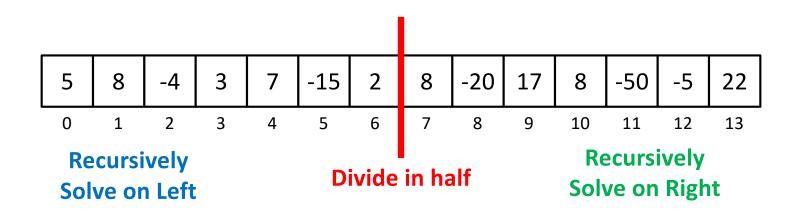


### Maximum Sum Continuous Subarray Problem

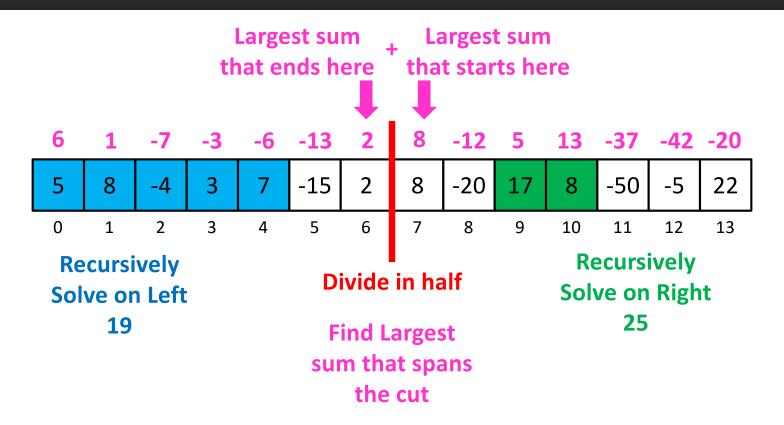
The maximum-sum subarray of a given array of integers A is the interval [a, b] such that the sum of all values in the array between a and b inclusive is maximal.

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 $\Theta(n \log n)$



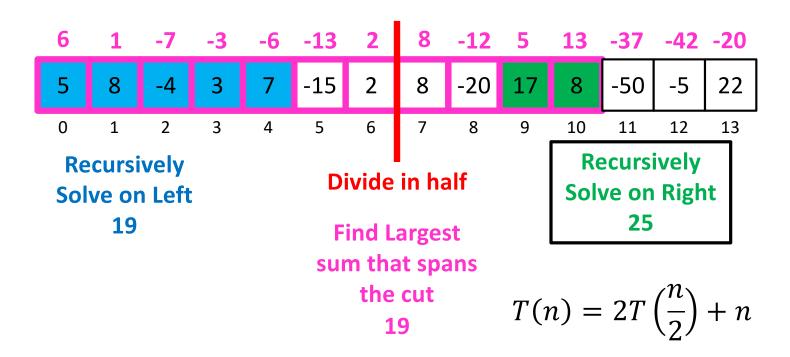
## Divide and Conquer $\Theta(n \log n)$



## 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</pre>
```

# 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

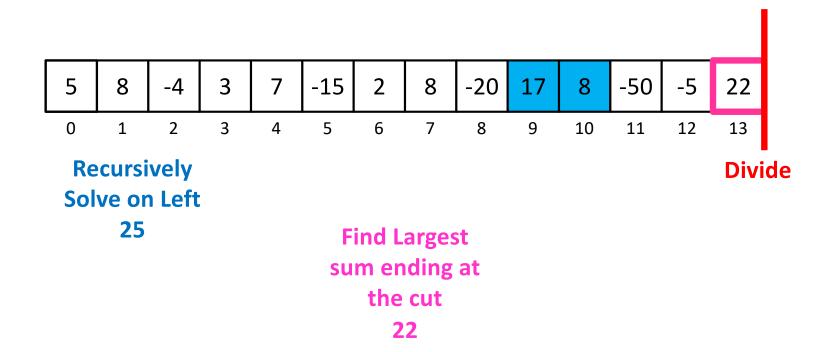
- Break the problem into a single smaller subproblem, recursively solve
- E.g. Impossible Missions Force (Double Agents), Quickselect, Binary
   Search

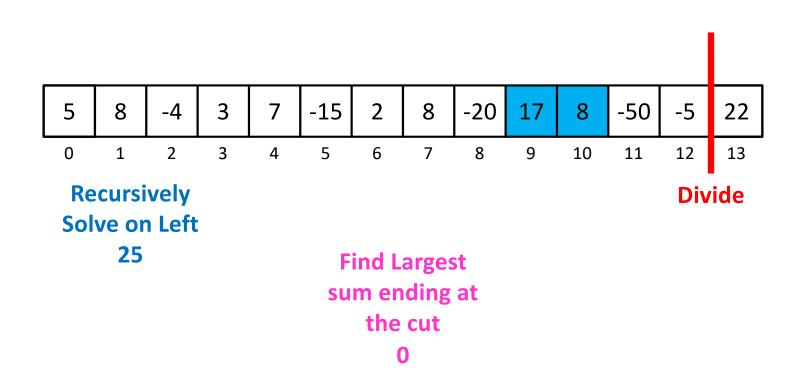
### Pattern So Far

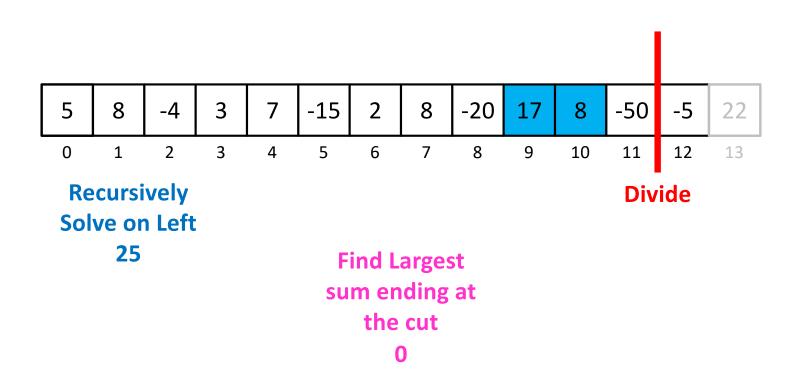
- Typically looking to divide the problem by some fraction (½, ¼ the size)
- Not necessarily always the best!
  - Sometimes, we can write faster algorithms by finding unbalanced divides.

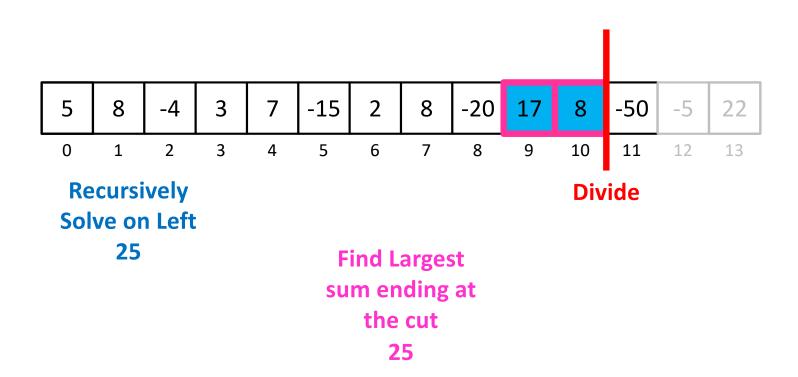
### Chip and Conquer

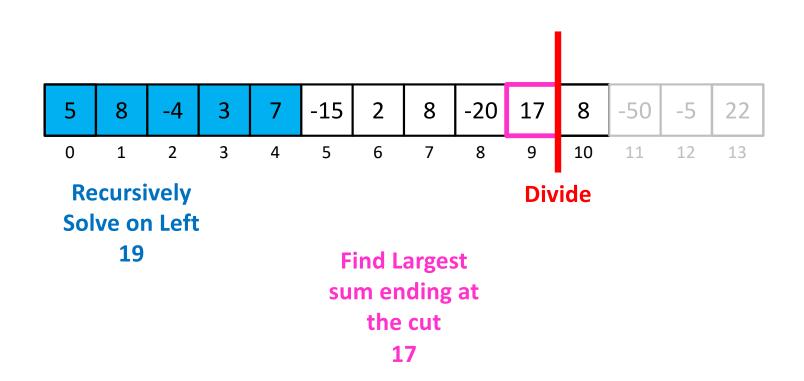
- Divide
  - Make a subproblem of all but the last element
- Conquer
  - Find best subarray on the left (BSL(n-1))
  - Find the best subarray ending at the divide (BED(n-1))
- Combine
  - New Best Ending at the Divide:
    - $BED(n) = \max(BED(n-1) + arr[n], 0)$
  - New best on the left:
    - $BSL(n)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

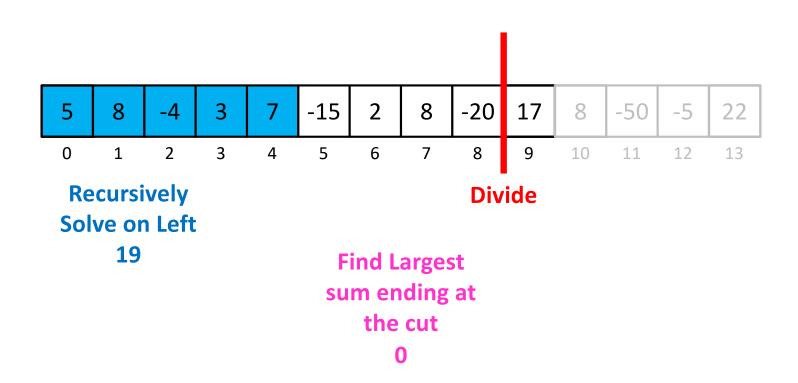


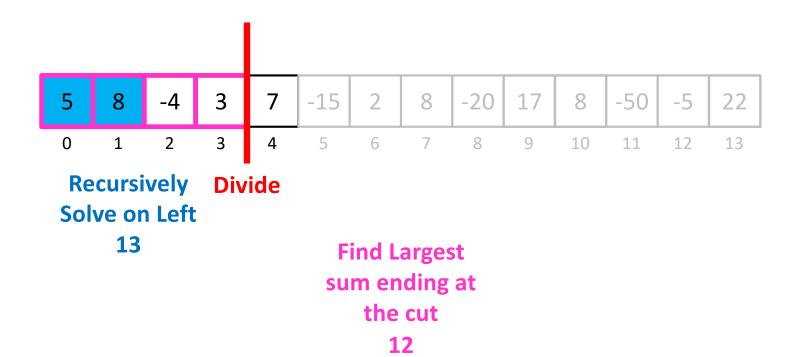












### Chip and Conquer

- 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

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

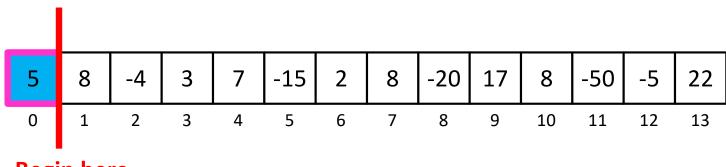
$$T(n) = \Theta(n \log n)$$

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

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

### MSCS Problem - Redux

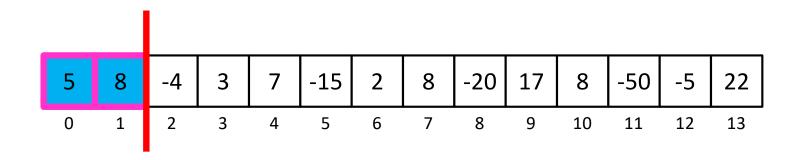
- Solve in O(n) by increasing the problem size by 1 each time.
- Idea: Only include negative values if the positives on both sides of it are "worth it"



**Begin here** 

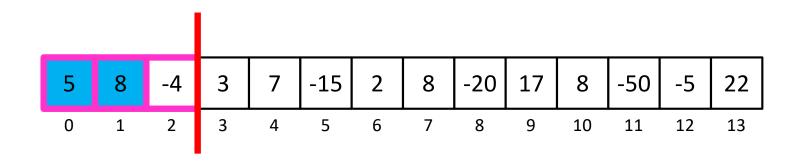
**Remember two values:** 

Best So Far 5



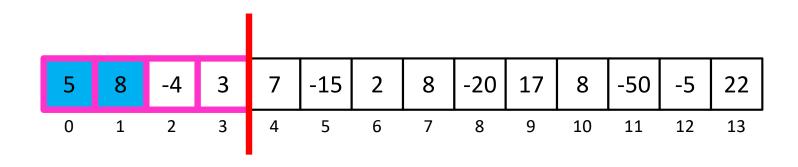
Remember two values:

Best So Far 13



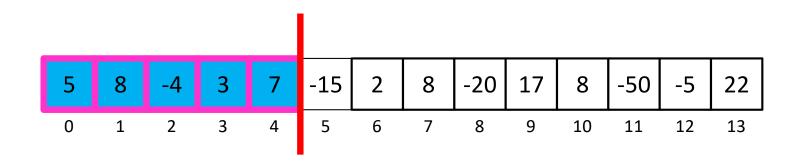
Remember two values:

Best So Far 13



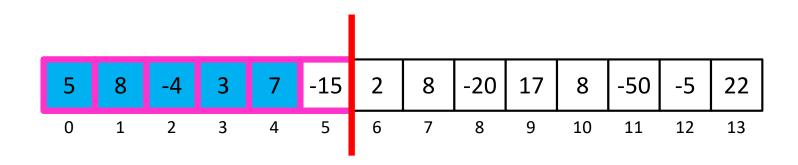
Remember two values:

Best So Far 13



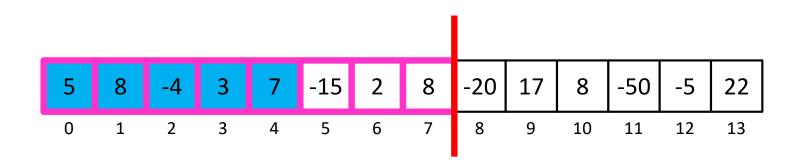
Remember two values:

Best So Far 19



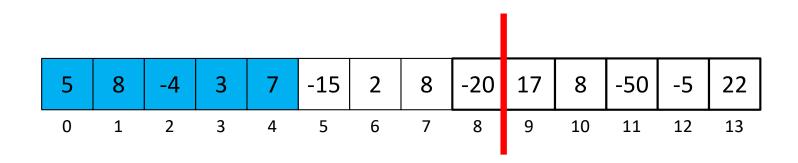
Remember two values:

Best So Far 19



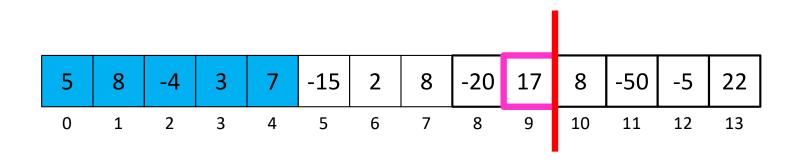
Remember two values:

Best So Far 19



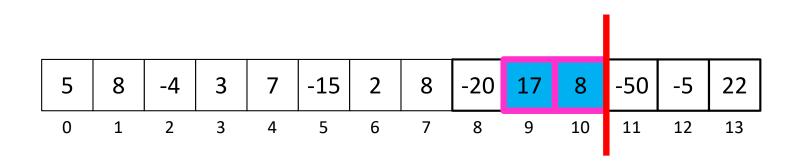
Remember two values:

Best So Far 19



Remember two values:

Best So Far 19



Remember two values:

Best So Far 25