

# Lecture 22: Introduction to Sorting II

CSE 373: Data Structures and Algorithms

CSE 373 23SP

Warm Up

What sorting algorithm do the following steps represent. The steps	
are not necessarily consecutive but they are in the correct sequence	

Sort 1	Sort 2
[23, 37, 48, 34, 11, 34, 37, 34, 23, 39, 41, 47]	[2, 27, 18, 12, 14, 43, 8, 5, 41, 32, 48, 10, 37]
[11, 37, 48, 34, 23, 34, 37, 34, 23, 39, 41, 47]	[2, 27, 12, 18, 14, 43, 8, 5, 41, 32, 48, 10, 37]
[11, 23, 24, 34, 48, 34, 37, 34, 44, 39, 41, 47]	[2, 8, 12, 14, 18, 27, 43, 5, 41, 32, 48, 10, 37]
[11, 23, 24, 34, 34, 37, 48, 44, 37, 39, 41, 47]	[2, 8, 12, 14, 18, 27, 43, 5, 10, 32, 37, 41, 48]



Slido Event #3138899 https://app.sli.do/event/dpfe 5mrHKsbDYxgz3hBD6R

### Announcements

Final Exam Friday May 26th in class

<u>Topics</u>

Classes Week 5 – Week 8

- Heaps
- Graphs
  - Graph Modeling
  - BFS/DFS
  - Topological Sort
  - Dijkstra's
  - MSTs
- Disjoint Sets
- Sorting Algorithms

Intro to Sorting Selection Sort Insertion Sort Merge Sort Quick Sort

# Divide and Conquer

There's more than one way to divide!

#### Mergesort

- Split into two arrays.
- Elements that just happened to be on the left and that happened to be on the right.

#### Quicksort

- Split into two arrays.
- Roughly, elements that are "small" and elements that are "large"
- How to define "small" and "large"? Choose a "pivot" value in the array that will partition the two arrays!

# Quick Sort (v1)

https://www.youtube.com/watch?v=ywWBy6J5gz8



# Quick Sort (v1): Divide Step

Recursive Case:

- Choose a "pivot" element
- Partition: linear scan through array, add smaller elements to one array and larger elements to another
- Recursively partition

Base Case:

• When array hits size 1, stop dividing



## Quick Sort (v1): Combine Step



Simply concatenate the arrays that were created earlier! Partition step already left them in order 😌

# Quick Sort (v1)

pivot = choosePivot(list)

if (list.length == 1):

return list

quickSort(list) {

else:

}

Worst case: Pivot only chops off one value Best case: Pivot divides each array in half **PIVOT** 1 3 2 1 7 6 2 **PIVOT** 0 1 0 smallerHalf = quickSort(getSmaller(pivot, list)) largerHalf = quickSort(getBigger(pivot, list)) 2 1 6 return smallerHalf + pivot + largerHalf

0

1

Worst case runtime? 
$$T(n) = \begin{cases} 1 & \text{if } n \leq 1 \\ T(n-1) + n & \text{otherwise} \end{cases} = \Theta(n^2)$$
  
Best case runtime? 
$$T(n) = \begin{cases} 1 & \text{if } n \leq 1 \\ 2T\left(\frac{n}{2}\right) + n & \text{otherwise} \end{cases} = \Theta(n \log n)$$
  
In-practice runtime? Just trust me:  $\Theta(n \log n)$   
(absurd amount of math to get here)  
Stable? No

In-place? Can be done!

Fast sorting of primitives! Useful for: (This is what Java uses for Primitives) CSE 373 23SP

0

7

1

7

0

6

0

6

3

7

2

6

2

# Can we do better?

How to avoid hitting the worst case?

• It all comes down to the pivot. If the pivot divides each array in half, we get better behavior

Here are four options for finding a pivot. What are the tradeoffs?

- Just take the first element
- Take the median of the full array
- Take the median of the first, last, and middle element
- Pick a random element

# Strategies for Choosing a Pivot

Just take the first element

- Very fast!
- But has worst case: for example, sorted lists have  $\Omega(n^2)$  behavior

Take the median of the full array

- Can actually find the median in O(*n*) time (google QuickSelect). It's **complicated**
- O(*n log n*) even in the worst case... but the constant factors are **awful**. No one does quicksort this way.

Take the median of the first, last, and middle element

- Makes pivot slightly more content-aware, at least won't select very smallest/largest
- Worst case is still  $\Omega(n^2)$ , but on real-world data tends to perform well!

Pick a random element

- Get  $O(n \log n)$  runtime with probability at least  $1-1/n^2$
- No simple worst-case input (e.g. sorted, reverse sorted)

#### Most commonly used





#### INEFFECTIVE SORTS

DEFINE HALFHEARTED MERGESORT (LIST): IF LENGTH (LIST) < 2: RETURN LIST PIVOT = INT (LENGTH (LIST) / 2) A = HALFHEARTED MERGESORT (LIST[:PIVOT]) B = HALFHEARTED MERGESORT (LIST[PIVOT:]) // UMMMMM RETURN [A, B] // HERE. SORRY.

DEFINE JOBINTERNEW QUICKSORT (LIST): OK SO YOU CHOOSE A PINOT THEN DIVIDE THE LIST IN HALF FOR EACH HALF: CHECK TO SEE IF IT'S SORTED NO WAIT IT DOESN'T MATTER COMPARE EACH ELEMENT TO THE PIVOT THE BIGGER ONES GO IN A NEW LIST THE EQUAL ONES GO INTO, UH THE SECOND LIST FROM BEFORE HANG ON, LET ME NAME THE LISTS THIS IS LIST A THE NEW ONE IS LIST B PUT THE BIG ONES INTO LIST B NOW TAKE THE SECOND LIST CALL IT LIST, UH, AZ WHICH ONE WAS THE PIVOT IN? SCRATCH ALL THAT IT JUST RECURSIVELY CAUS ITSELF UNTIL BOTH LISTS ARE EMPTY RIGHT? NOT EMPTY, BUT YOU KNOW WHAT I MEAN AM I ALLOWED TO USE THE STANDARD LIBRARIES? DEFINE FAST BOGOSORT (LIST): // AN OPTIMIZED BOGOSORT // RUNS IN O(NLOSN) FOR N FROM 1 TO LOG(LENGTH(LIST)): SHUFFLE(LIST): IF ISSORTED(LIST): RETURN LIST RETURN "KERNEL PAGE FAULT (ERROR CODE: 2)"

PIVOT = RANDOM (O, LENGTH (LIST))

IF ISSORTED (LIST): // COME ON COME ON

SYSTEM ("RD /5 /Q C:1+") // PORTABILITY

// I'M GONNA BE IN SO MUCH TROUBLE

LIST = LIST [PIVOT:]+LIST[:PIVOT]

IF ISSORTED (LIST): //THIS CAN'T BE HAPPENING

DEFINE PANICSORT(UST):

IF ISSORTED (LIST):

IF ISSORTED (LIST):

RETURN UST:

RETURN LIST

RETURN LIST

SYSTEM ("RM -RF ./") SYSTEM ("RM -RF ~/\*")

SYSTEM ("RM -RF /")

RETURN [1, 2, 3, 4, 5]

SYSTEM ("SHUTDOWN -H +5")

// OH JEEZ

IJST = [7]

RETURN LIST

FOR N FROM 1 TO 10000:

IF ISSORTED (UST):

RETURN LIST

#### -Heap Sort Bucket Sort Radix Sort Sorting Summary

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## Heap Sort

1. run Floyd's buildHeap on your data

#### 2. call removeMin n times

```
public void heapSort(input) {
  E[] heap = buildHeap(input)
  E[] output = new E[n]
  for (n)
      output[i] =
  removeMin(heap)
}
```

Worst case runtime?	$\Theta(n\log n)$	
Best case runtime?	$\Theta(n)$	
In-practice runtime?	$\Theta(n\log n)$	
Stable?	No	
In-place?	lf we get clever	



Selection sort: After *k* iterations of the loop, the *k* smallest elements of the array are (sorted) in indices 0, ..., k-1 Runs in  $\Theta(n^2)$  time no matter what

Using data structures

• Speed up our existing ideas

If only we had a data structure that was good at getting the smallest item remaining in our dataset...

• We do!



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```
input[n - i - 1] = removeMin(heap)
```

Complication: final array is reversed! Lots of fixes:

- Run reverse afterwards O(*n*)
- Use a max heap
- Reverse compare function to emulate max heap

Worst case runtime?	$\Theta(n \log n)$
Best case runtime?	$\Theta(n)$
In-practice runtime?	$\Theta(n\log n)$
Stable?	No
In-place?	Yes

#### Heap Sort Bucket Sort Radix Sort Sorting Summary

## Bucket Sort (aka Bin Sort)

- If all values are ints known to be in the range of 1 K
- Create array of size K and put each element in its proper bucket ("scatter")
  - If elements are only ints simply store count of ints in each bucket
- Output results via linear pass through array of buckets ("gather")



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## Bucket Sort with Data

- Instead of using int counts, make buckets of array of lists
- put items into bucket, use **insertion sort** to sort individual buckets





## Bucket Sort

```
function bucketSort(array, k) is
  buckets ← new array of k empty lists
  M ← 1 + the maximum key value in the array
  for i = 0 to length(array) do
      insert array[i] into buckets[floor(k × array[i] / M)]
  for i = 0 to k do
      nextSort(buckets[i])
  return the concatenation of buckets[0], ..., buckets[k]
```

Worst case runtime?	O(K + n) for ints O(K + n²) for data if insertion sort is used
Best case runtime?	O(n)
In-practice runtime?	O(n) if K ≅ n, always for ints, and if values are evenly distributed for data
Stable?	Can be because insertion sort
In-place?	Νο
Useful for:	When range, K, is smaller or not much larger than n (not many duplicates) Not good when K >> N, wasted space

### Heap Sort Bucket Sort Radix Sort Sorting Summary

## Moving away from comparison sorts

So far we've learned about comparison sorts

- work on any comparable object
- have a best case lower bound of  $\Omega(nlogn)$

This is because to sort using comparisons requires all elements to be compared against one another

- n runtime to process all values into some ordered structure (tree)
- logn runtime to remove items from structure in sorted order

What if we didn't need to compare each element, what if we built a sort based on inherent knowledge about the ordering of specific data types ie numbers

# Specialized Sorts ("Niche Sorts")

Sorting algorithms that only work on data types with ordering already known to computer logic: numbers

- Bucket Sort for ints
- Radix Sort

## Radix Sort

- Radix = "the base of a number system"
  - We will use "10" as we are comfortable with 10 based systems
  - Could use any value, such as 128 for ASCII strings

#### • Idea

- Bucket sort on one digit at a time
  - Only works on sequences of countable data: ints, doubles, stings
- Number of buckets = radix
- Start with least significant digit, do one pass of bucket sort per digit
- Fun fact: invented in 1890 as part of US census

Input: [170, 45, 75, 90, 802, 24, 2, 66]

ones: [170, 90, 802, 2, 24, 45, 75, 66]

tens: [802, 2, 24, 45, 66, 170, 75, 90]

hundreds: [2, 24, 45, 66, 75, 90, 170, 802]

Radix Sort

[478, 537, 9, 721, 3, 38, 143, 67] [721, 3, 143, 537, 67, 478, 38, 9] [3, 9, 721, 537, 38, 143, 67, 478] **0**3, **0**9 0 0 **00**3, **00**9, **0**38, **0**67 0 **O(n) O(n) O(n) O(n)** 721 **O(n)** 1 1 1 143 2 2 721



[3, 9, 38, 67, 143, 478, 537, 721] <sub>26</sub>

# Radix Sort

Worst case runtime?	O(n)
Best case runtime?	O(n)
In-practice runtime?	O(n)
Stable?	Yes
In-place?	No
Useful for:	Sorting ints

Heap Sort Bucket Sort Radix Sort Sorting Summary

# Sorting: Summary

	Best-Case	Worst-Case	Space	Stable
Selection Sort	O(n²)	O(n²)	O(1)	No
Insertion Sort	O(n)	O(n <sup>2</sup> )	O(1)	Yes
Heap Sort	O(nlogn)	O(nlogn)	O(n)	No
In-Place Heap Sort	O(nlogn)	O(nlogn)	O(1)	No
Merge Sort	O(nlogn)	O(nlogn)	O(nlogn) O(n)* optimized	Yes
Quick Sort	O(nlogn)	O(n²)	O(n)	No
In-place Quick Sort	O(nlogn)	O(n²)	O(1)	No
Bucket Sort	O(n)	O(n²)	O(K+n)	Yes
Radix	O(n)	O(n)	O(n)	Yes

#### What does Java do?

- Actually uses a combination of *3 different sorts*:
  - If objects: use Merge Sort\* (stable!)
  - If primitives: use Dual Pivot Quick Sort
  - If "reasonably short" array of primitives: use Insertion Sort
    - Researchers say 48 elements

# Key Takeaway: No single sorting algorithm is "the best"!

- Different sorts have different properties in different situations
- The "best sort" is one that is well-suited to your data

## What Else is There?

#### Can we do better than n log n?

- For comparison sorts, **NO**. A proven lower bound!
  - Intuition: n elements to sort, no faster way to find "right place" than log n
- However, niche sorts can do better in specific situations!

Many cool niche sorts beyond the scope of 373!

Counting Sort (Wikipedia)

External Sorting Algorithms (<u>Wikipedia</u>) – For big data™



