

# More Sorting

# Quicksort

- Quicksort is very commonly used, efficient, comparison based sorting algorithm.
- It is a **divide and conquer** algorithm.
- Not stable.
  
- Its average case time complexity is  $O(N \log N)$ ,
  - which is the same as Mergesort
  - but its constants are typically better than Mergesort
  - Hence it usually performs better than Mergsort.

# Quicksort

**QuickSort** ( A )

**QSort** ( A, 0, A.length – 1 );

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**QSort** ( A, start, end )

    if ( start == end )

        return;

    pivotIndex = **PickAPivotIndex** ( start, end )

    pivotFinalIndex = **Partition** ( A, start, end, pivotIndex )      // Now the element at *pivotFinalIndex* is at its final sorted position.

**QSort**( A, start, pivotFinalIndex – 1 );      // sort sub array to left of *pivotFinalIndex*

**QSort**( A, pivotFinalIndex + 1, end );      // sort sub array to right of *pivotFinalIndex*

# Quicksort

1<sup>st</sup> row shows state after partition has completed

start				leftMostBigger					end
16	4	7	3	17	34	25	52	88	43

**Elements less than *pivot***

**Elements greater than or equal to *pivot***

# Quicksort

2<sup>nd</sup> row below shows state after swap happens ( Note: swap happens after partition completes )

start				leftMostBigger					end
16	4	7	3	17	34	25	52	88	43
3	4	7	16	17	34	25	52	88	43

*Elements less than pivot*

*Elements greater than or equal to pivot*

Note that pivot element 16 is now in its final place in the sorted array

# Quicksort

start				leftMostBigger					end
3	4	7	16	17	34	25	52	88	43

Now, we will call QSort on

- sub array to left of **partition element**
- sub array to right of **partition element**
- and we won't touch the **partition element**

# Quicksort

**PickAPivotIndex** (start, end )

return random number between start and end ( inclusive ).

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// move all elements less than pivot value to left of pivot, and all elements greater than pivot value to its right.

**Partition** ( A, start, end, pivotIndex )

Swap ( A[ start ], A [ pivotIndex ] );

newStart = start + 1;

leftMostBiggerIndex = newStart;

for ( ii = newStart; ii <= end; ++ ii )

if A [ ii ] < A [ pivotIndex ]

Swap ( A [ ii ], A [ leftMostBiggerIndex ] );

++ leftMostBiggerIndex;

pivotFinalIndex = leftMostBiggerIndex - 1;

// final position of pivot is immediate left of leftMostBiggerIndex

Swap ( A[ start ], A[ pivotFinalIndex ] );

// put pivot in its final position

return pivotFinalIndex;

# Quicksort

- In place
- Not stable
- Time complexity
  - Best case and average :  $O(N \log N)$
  - Worst case:  $O(N^2)$ 
    - Worst case can happen:
      1. if pivots are consistently smallest or largest between *start* and *end*.  
However, this is very hard to achieve 😊 and probability of that happening is extremely low.  
See next slide for details.
      2. If there are a lot of duplicates.  
In this case, use 3 way partitioning. (Look up the Dutch national flag problem).
- Space complexity:  $O(\log N)$  due to recursion.



# Quicksort

- One way you would get the worst case time:
  - Take a sorted or almost sorted array (or reverse sorted).
  - Always pick the first or last element between *start* and *end* as the pivot.
  - This means u will partition the array with one element on one side and rest of the array on other side.
  - To prevent against this, pick a random pivot.
  - Note: You could also pick the middle element and be OK, but input can be crafted against this methodology, so better to pick random.

# Quicksort

- Cache locality
  - Quicksort has good cache locality.
- Its better than Merge sort (which also accesses consecutive elements in an array) because Mergesort uses an auxiliary array.

## LAB 1 of 2

- Write a function that sorts the input array using the quick sort algorithm.
  - For picking the pivot:
    - If u know a random number generation function, can use it.
    - If not, just pick middle element (from start and end ) for now.

# Quicksort

Time complexity recurrence relation

**Best case:**

$$T(N) = 2 T(N/2) + cN$$

$$T(N/2) = 2 T(N/4) + cN/2$$

$$T(N/4) = 2 T(N/8) + cN/4$$

and so on...

← Assumes we partition the array into **half** each time ( unreasonable assumption 😊 )

Above comes to  $T(N) = N \log N$

# Quicksort

## Time complexity recurrence relation

### **Worst case:**

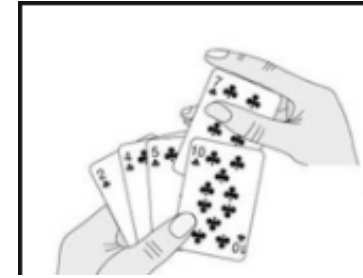
← Every time, we reduce the array by only 1 element (pick the biggest or smallest element each time).

$$\begin{aligned}T(N) &= T(N-1) + cN \\&= T(N-2) + c(N-1) + cN \\&= T(1) + c2 + c3 + \dots + c(N-1) + cN \\&= c1 + c2 + c3 + \dots + c(N-1) + cN \\&= c(1 + 2 + 3 + \dots + (N-1) + N) \\&= c(N(N+1)/2) \\&= O(N^2)\end{aligned}$$

# Insertion sort

- Its not very efficient in general.
- It's a great choice if, say, the input is already sorted and u want to add a new element to it...  $O(N)$
- In place
- Stable
- Time complexity
  - Worst and average case:  $O(N^2)$
  - Best case:  $O(N)$

# Insertion sort



- Lets think of this algo in the following way:
  - You have to sort  $N$  numbers
  - Assume the first  $N-1$  numbers are somehow already sorted.
  - Now, the problem is just to put the  $N^{\text{th}}$  number in the right place inside sorted array with  $N-1$  elements.
  - Extend that thinking to:
    - assume first  $N-2$  numbers already sorted.
    - Now u need to just put the  $(N-1)^{\text{th}}$  and  $N^{\text{th}}$  number in the right place inside this sorted array.
    - And so on...
- Here we are thinking of a top down approach (recursive).
- Another approach would be bottom up (iterative)

# Insertion sort - pseudocode for recursive

**InsertionSort ( A )**

    InSort( A, A.length – 1 );

**InSort( A, N )**

    if ( N <= 1 )

        return;

**InSort( A, N – 1 );**                      // first lets sort N-1 numbers

    // If we r here, it means we sorted N – 1 numbers already ( and the first time we will come here is when N has a value of 2, so N-1 is just 1 )

    // remember, N here is NOT ALWAYS A.length-1, we are recursing, so N is whatever it is on call stack here

    i = N;

    // We look for a place to insert A [ N ]

    while ( i > 0 and A[ i ] < A [ i – 1 ] )

    // While the element is less than its immediate left neighbor

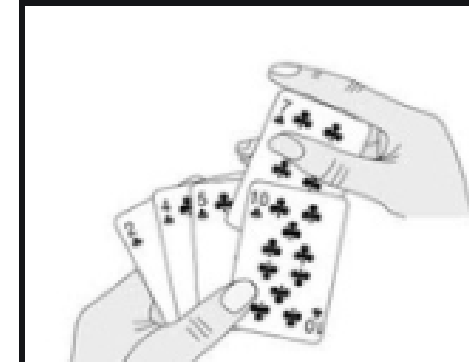
        Swap ( A[ i ], A [ i – 1 ] );

        --i;



# Insertion sort

- Iterative approach
  - I can solve the problem for one element.
  - Then solve for two elements.
  - ...
  - Then solve for  $N$  elements



# Insertion sort – pseudocode for Iterative

**InsertionSort ( A )**

    InSort( A, A.length – 1 );

**InSort( A, N )**

    if ( N <= 1 )

        return;

    for i = 1 to N               // start from index 1, think of element at 0<sup>th</sup> index as a sorted sub array of size 1.

        j = i – 1;

        while ( j >= 0 and A [ j+1 ] < A [ j ] )               // While the element is less than its immediate left neighbor

            Swap ( A[ j + 1 ], A [ j ] );

            --j;

QQ: What implications if u change the comparison to be <=

    while ( j >= 0 and A [ j+1 ] <= A [ j ] )

        Swap ( A[ j + 1 ], A [ j ] );

        --j;

# Insertion sort

Example:

• Input:	16	4	7 <sub>a</sub>	3	17	34	25	7 <sub>b</sub>	
•	16	4	7 <sub>a</sub>	3	17	34	25	7 <sub>b</sub>	
•	4	16	7 <sub>a</sub>	3	17	34	25	7 <sub>b</sub>	
•	4	7 <sub>a</sub>	16	3	17	34	25	7 <sub>b</sub>	
•	3	4	7 <sub>a</sub>	16	17	34	25	7 <sub>b</sub>	
•	3	4	7 <sub>a</sub>	16	17	34	25	7 <sub>b</sub>	
•	3	4	7 <sub>a</sub>	16	17	34	25	7 <sub>b</sub>	
•	3	4	7 <sub>a</sub>	16	17	25	34	7 <sub>b</sub>	
•	3	4	7 <sub>a</sub>	7 <sub>b</sub>	16	17	25	34	← Sorted, and stable

# Insertion sort

- Lets look at the **best case** ... input is already sorted (or almost)

•	4	7 <sub>a</sub>	7 <sub>b</sub>	16	17	25	34	← Starting input
•	4	7 <sub>a</sub>	7 <sub>b</sub>	16	17	25	34	
•	4	7 <sub>a</sub>	7 <sub>b</sub>	16	17	25	34	← No movement here or in any row below
•	4	7 <sub>a</sub>	7 <sub>b</sub>	16	17	25	34	
•	4	7 <sub>a</sub>	7 <sub>b</sub>	16	17	25	34	
•	4	7 <sub>a</sub>	7 <sub>b</sub>	16	17	25	34	
•	4	7 <sub>a</sub>	7 <sub>b</sub>	16	17	25	34	
•	4	7 <sub>a</sub>	7 <sub>b</sub>	16	17	25	34	

- First element is compared with its immediate left neighbor, and we decide nothing needs to be done.
- Move to next element, again compare with its immediate left neighbor, nothing needs to be done.
- So, in each iteration, there is one comparison and no movement....  $O(1)$
- Doing this for  $N-1$  elements ...  $O(N)$  is the complexity for this best case scenario

# Insertion sort

- Now let's look at the **worst case** scenario ... input is reverse sorted.
  - Every element in the starting input is less than ALL the elements to its left.

- 34    25    17    16    7<sub>a</sub>    7<sub>b</sub>    4
- 34    25    17    16    7<sub>a</sub>    7<sub>b</sub>    4    ← 25: 1 comparison
- 25    34    17    16    7<sub>a</sub>    7<sub>b</sub>    4    ← 17: < all elements to its left. 2 comparisons
- 17    25    34    16    7<sub>a</sub>    7<sub>b</sub>    4    ← 16: 3 comparisons for moving 16
- 16    17    25    34    7<sub>a</sub>    7<sub>b</sub>    4
- 7<sub>a</sub>    16    17    25    34    7<sub>b</sub>    4
- 7<sub>a</sub>    7<sub>b</sub>    16    17    25    34    4    ← 4: 6 comparisons → N-1 comparisons
- 4    7<sub>a</sub>    7<sub>b</sub>    16    17    25    34    ← 4 had to move all the way to the beginning

- Total comparisons: 1 + 2 + ... + (N-1) = O(N<sup>2</sup>)

# Counting sort

- Counting sort is not a comparison based algorithm.
- It is not a general purpose sorting algorithm... values have to be integers within a known “small” range.
- Not in place
- Usually linear time and space complexity.

## Counting sort

- Input:            2            4            7            3            1            2            0            7            7

index	0	1	2	3	4	5	6	7
Counts / frequency	1	1	2	1	1	0	0	3

- We use a counting array, whose size is K where K is the maximum value in the input    ← Space is  $O(K)$
- For each element x in input    ←  $O(N)$ 
  - ++ countingArray[ x ]    ←  $O(1)$
- For each index ii in the counting array    ←  $O(K)$ 
  - If ( countingArray [ ii ] > 0 )
    - print ii            // number of times we print is countingArray [ ii ]

# Counting sort

- Input: 2      4      7      3      1      2      0      7      7

index	0	1	2	3	4	5	6	7
Counts / frequency	1	1	2	1	1	0	0	3

- We use a counting array, whose size is K where K is the maximum value in the input ← Space is  $O(K)$
- Initialize countingArray to all 0s ←  $O(K)$
- For each element x in input  
  ++ countingArray[ x ] ←  $O(N)$   
                                  ←  $O(1)$
- For each index ii in the counting array ←  $O(K)$   
  If ( countingArray [ ii ] > 0 )  
    print ii      // number of times we print is countingArray [ ii ]
- Complexity
  - Time:  $O(N + K)$ .
  - Space:  $O(K)$ . ← we will see how this might change a bit due to counting sort on objects instead of integers.
  - If K is “small enough”, then this is linear complexity (this is when this algo would be / could be used).
  - If K is, say,  $N^2$ , then this is not linear time complexity, or linear space complexity.
- Next slide:** What if we have objects as input, instead of just numbers.



## Counting sort -- Objects

- If we are dealing with objects (common scenario) instead of numbers:
  - We would need to store the objects in the counting array.
  - And Count sort would still need integers to sort
    - Some integer attribute of the objects (or something that can be mapped to an integer value)
  - And to keep the sort stable, we would keep the “order of insertion” in mind.
    - “order of insertion” refers to storing more than one object having the same sorting criteria.

Index	0	1	2	...	K-1
	Object A	Object B1		Object C	Object D1
		Object B2			Object D2
					Object D3

- Space complexity now is:
  - $O(K + N)$  since we have a counting array of size K and we store N objects in it.

# Choosing a sort algo

- It depends on what is needed in ur solution to the problem u r trying to solve.
- If u care about stability: Merge Sort.
  - Trade off ?
- If u cannot use extra space: Quicksort
  - Trade off?
- If u have a “small” sized input or almost sorted data: Insertion Sort
- If input is in a restricted range: Count sort
- When would u use Heapsort?
  - Pros: Guaranteed  $O(N \log N)$  time.  $O(1)$  space
  - Cons: Not stable. Bad cache locality

# Sorting cheat sheet

Algo	Worst	Avg	Best	Stable	InPlace	Space	Other
Bubble	$N^2$	$N^2$	$N^2$	Yes	Yes	$O(1)$	Comparison sort. Stable bcos same element wont jump over. E.g.: 3 9 7a 11 17 7b Here, 7b will not go to the left of 7a, so relative ordering maintained.
Insertion	$N^2$	$N^2$	N	Yes	Yes	$O(1)$	Comparison sort. Best is N bcos no shifting or almost no shifting is required when array is sorted or almost sorted. Worst case is when reverse sorted. Running time can change depending on values and position of values (unlike Bubble or Selection sort)
Selection	$N^2$	$N^2$	$N^2$	No (see comments)	Yes	$O(1)$	Comparison sort. Not stable because this can happen: 7a 3 9 7b 1 Find min, which is 1, and swapping with first elem gives 1 3 9 7b 7a. So order of 7a and 7b is switched, making it unstable. To make it stable, can insert 1 in its <b>place</b> , instead of swapping (but that is expensive due to shifting). 1 7a 3 9 7b
Quicksort	$N^2$	$N \log N$	$N \log N$	No	Yes	$O(\log N)$	Comparison sort. QS is in place, but does use stack space for recursion. Standard QS not stable, but can use extra mem to get a stable version, but then its not in place. Standard implementation is in place (not stable). Its constants are usually better than MS (merge sort), hence it is usually faster than MS. Worst case can happen if array is already sorted (or reverse sorted) or almost sorted, and u pick first or last element as pivot. If u always pick the middle element, input can be crafted to cause worst case behavior. Just pick random element. Note, if array is randomly ordered (not sorted), then picking first or last element as pivot would be ok, but its better to pick a random element (or median) as pivot to guard against the case where array is already sorted or almost sorted. QS can be implemented <a href="#">iteratively</a> if u use ur own stack. Cache locality of QS is good. Often it is better than cache locality for MS, because MS uses a 2nd array in addition to input array.
Mergesort	$N \log N$	$N \log N$	$N \log N$	Yes	No	$O(N)$	Comparison sort. Run time is predictable, always $O(N \log N)$ regardless of input values/position. Cache locality is good, but QS generally has slightly better cache locality since MS uses an additional array (its not in place)
Heapsort	$N \log N$	$N \log N$	$N \log N$	No	Yes	$O(N)$	Comparison sort. Cache locality not great due to all the long range swapping within the heap. Guaranteed $O(N \log N)$ time, $O(1)$ space
Radixsort	$O(nK)$	$O(nK)$	$O(nK)$	Yes	No	$O(N+K)$	
Counting sort				Yes	No	$O(K)$	Not in place as it uses a frequency array
Bucket sort						$O(N)$	
Tim sort	$N \log N$	$N \log N$	N	yes		$O(N)$	Hybrid of merge and insertion. Used in Python. Used in Java to sort objects (because stability could be important when sorting objects due to objects potentially having metadata/satellite data. But in Java if u r sorting primitive types, stability is not important and it uses Quicksort since that is typically faster than Mergesort.
Introsort				No	Yes		Comparison sort. Hybrid of QS, heap and insertion. Starts with QS, if recursion depth exceeds some limit, switches to heap sort, and if number of elements falls below a threshold, switches to insertion sort. C++ sort <i>typically</i> uses Introsort (could vary depending on implementation of C++ STL). C++ stable_sort typically uses a version of mergesort.

## LAB 2 of 2

- Take the same input you used in the quicksort lab (just so u can compare the two algorithm run times)
- Implement insertion sort
- Do u see a difference in the run times of quicksort and insertion sort when running on the same input?