



## CS202 – Data Structures

**LECTURE-24**

# Sorting – II

More on Sorting algorithms

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# Agenda

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- Sorting Algorithms
  - Quick Sort
  - External Sort
  - Linear Sorting Algorithms

# Quicksort

# Quick Sort

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- Choose a “pivot” element
- Group your elements into three groups:
  - Less than pivot
  - Equal to pivot
  - Greater than pivot
- Recursively sort (quick sort) the less than and greater than groups
- Concatenate the three sorted groups back together again

# Algorithm

1. Start with list  $I$  of  $n$  items
2. Choose a **pivot item  $v$**  from  $I$
3. Partition  $I$  into 2 unsorted lists  $I_1$  and  $I_2$  around the pivot  $v$ 
  - $I_1$ : All keys smaller than  $v$
  - $I_2$ : All keys larger than  $v$
  - Items equal to  $v$  can go in either list and pivot is not part of  $I_1$  or  $I_2$
4. Sort  $I_1$  and  $I_2$  recursively, yielding sorted lists  $S_1$  and  $S_2$
5. Concatenate  $S_1$ ,  $v$ ,  $S_2$  yielding the sorted list  $S$

**Base case:** list of size 0 or 1 (already sorted!)

# Quicksort Example

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5	3	9	4	8	2	1	6
---	---	---	---	---	---	---	---

*Suppose we pick the first element (i.e., 5) as the pivot*

# Another Example

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0	1	3	4	5	7	9
---	---	---	---	---	---	---

*Suppose we pick the first element (i.e., 0) as the pivot*

What will be the resulting complexity?

# Given an array of integers, which number/integer should be selected as a pivot?

Nobody has responded yet.

Hang tight! Responses are coming in.

# What are some ways to pick the pivot?

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- Always pick the **first element**
- Randomly pick an element from the list (**Randomized QS**)
- Randomly pick 3 elements and then choose the median of these 3 elements as the pivot (**median-of-3 strategy**)
- Note: With 2 and 3, the expected running time is in  $\Theta(n \log n)$

# What is the ideal pivot?

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- Ideal pivot: median

How can we efficiently find the Median?

# Pivot Selection

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How can we efficiently find the Median?

# How can we efficiently find the median?

We can find the median in  $O(n)$  time using **BFPRT** (called **PICK** in original paper)

JOURNAL OF COMPUTER AND SYSTEM SCIENCES 7, 448–461 (1973)

**Historical note:** The authors of this paper include **FOUR** Turing Award winners

## Time Bounds for Selection\*

MANUEL BLUM, ROBERT W. FLOYD, VAUGHAN PRATT,  
RONALD L. RIVEST, AND ROBERT E. TARJAN

*Department of Computer Science, Stanford University, Stanford, California 94305*

Received November 14, 1972

The number of comparisons required to select the  $i$ -th smallest of  $n$  numbers is shown to be at most a linear function of  $n$  by analysis of a new selection algorithm—PICK. Specifically, no more than  $5.4305 n$  comparisons are ever required. This bound is improved for extreme values of  $i$ , and a new lower bound on the requisite number

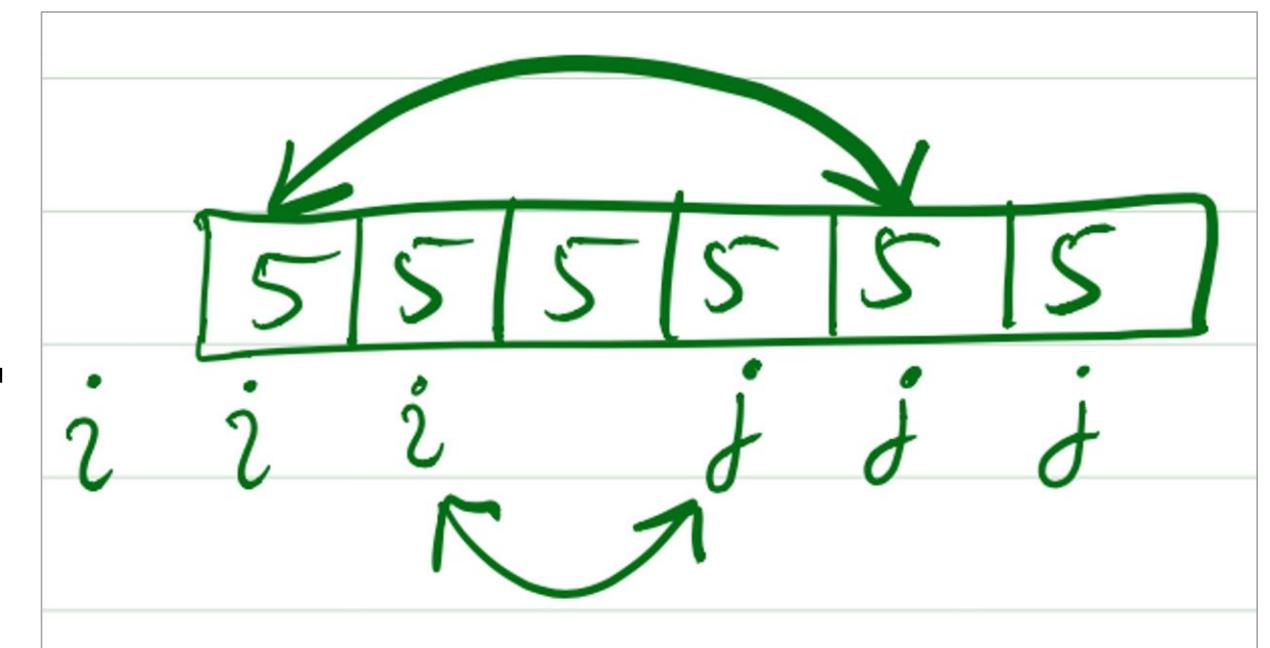
# Quicksort if input is ...

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- If we pick the pivot randomly
  - Array:  $\Theta(1)$  to read the value
  - Linked List:  $\Theta(n)$  to read the random pivot

# Invariants & Duplicate Values

- All items left of  $i$  are less than the pivot
- All items right of  $i$  are greater than the pivot
- Handling keys equal to the pivot with arrays
  - The solution is to make sure each index,  $i$  and  $j$ , stops whenever it reaches a key equal to the pivot and participates in a swap
  - If all items have the same key, half go into  $I_1$  and half into  $I_2$ , well-balanced recursion tree,  $O(n \log n)$  time



```

void quicksort(int a[], int low, int high) {
    // If there are fewer than two items, do nothing.
    if (low < high) {

        // Generate a random number in between low .. high
        srand(time(NULL));
        int pivotIndex = low + rand() % (high - low);
        int pivot = a[pivotIndex];

        // Swap pivot with last item (since this is in-place)
        a[pivotIndex] = a[high];
        a[high] = pivot;

        int i = low - 1;
        int j = high;
        int tmp; // temporary variable for swapping

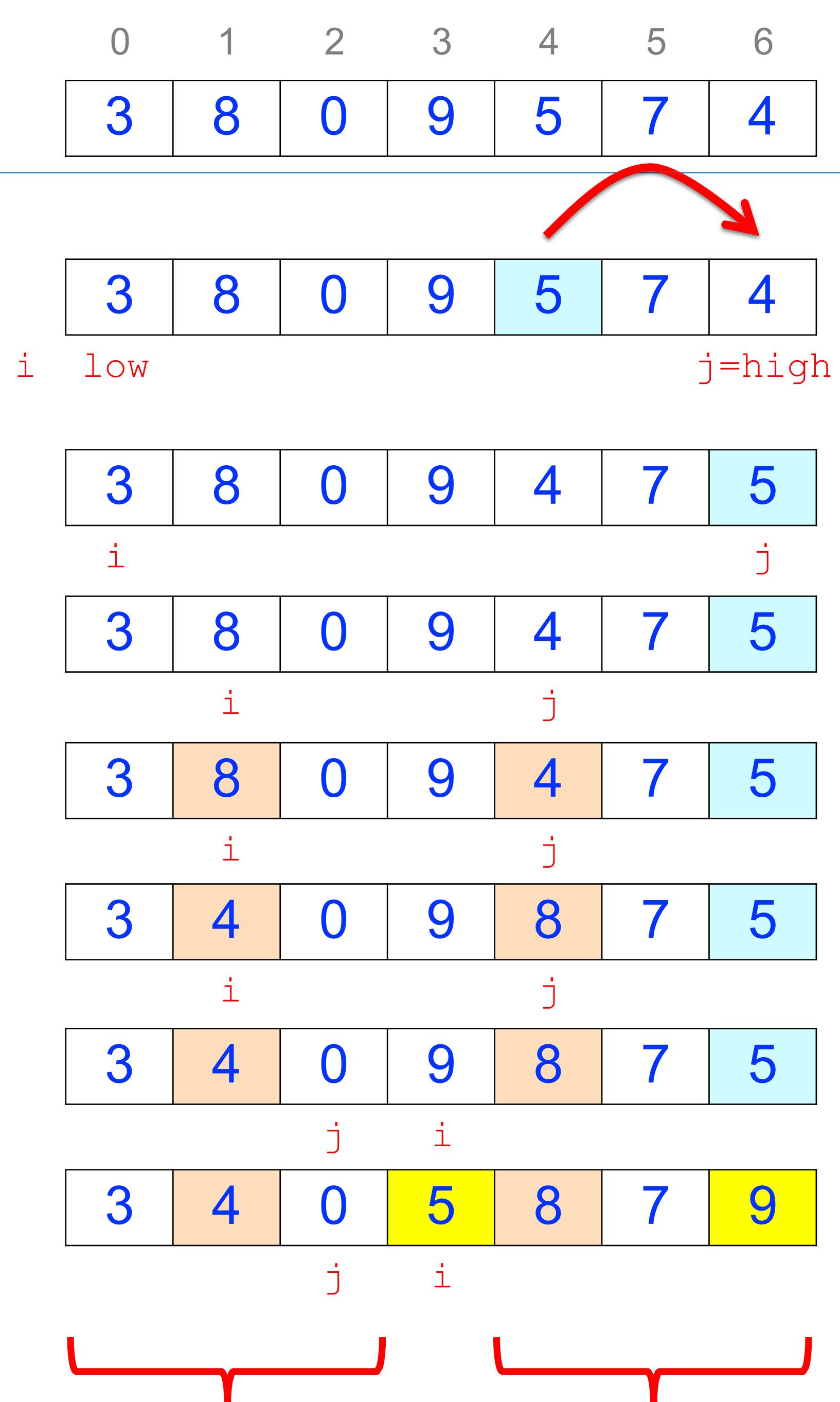
        do {
            do { i++; } while (a[i] < pivot);
            do { j--; } while ((a[j] > pivot) && (j > low));
            if (i < j) {
                // swap a[i] and a[j]
                tmp = a[i];
                a[i] = a[j];
                a[j] = tmp;
            }
        } while (i < j);

        a[high] = a[i];
        a[i] = pivot; // Put pivot in the middle where it belongs

        cout<<"pivot = "<< pivot << ", (low,high) = (" << low << "," << high << "): ";
        printArray(a, array_size);

        quicksort(a, low, i - 1); // Recursively sort left list
        quicksort(a, i + 1, high); // Recursively sort right list
    }
}

```



# Quick Sort

**Algorithm** inPlaceQuickSort( $S, a, b$ ):

**Input:** An array  $S$  of distinct elements; integers  $a$  and  $b$

**Output:** Array  $S$  with elements originally from indices from  $a$  to  $b$ , inclusive,  
sorted in nondecreasing order from indices  $a$  to  $b$

**if**  $a \geq b$  **then return** {at most one element in subrange}

$p \leftarrow S[b]$  {the pivot}

$l \leftarrow a$  {will scan rightward}

$r \leftarrow b - 1$  {will scan leftward}

**while**  $l \leq r$  **do**

{find an element larger than the pivot}

**while**  $l \leq r$  and  $S[l] \leq p$  **do**

$l \leftarrow l + 1$

{find an element smaller than the pivot}

**while**  $r \geq l$  and  $S[r] \geq p$  **do**

$r \leftarrow r - 1$

**if**  $l < r$  **then**

swap the elements at  $S[l]$  and  $S[r]$

{put the pivot into its final place}

swap the elements at  $S[l]$  and  $S[b]$

{recursive calls}

inPlaceQuickSort( $S, a, l - 1$ )

inPlaceQuickSort( $S, l + 1, b$ )

{we are done at this point, since the sorted subarrays are already consecutive}

**Code Fragment 11.6:** In-place quick-sort for an input array  $S$ .

# Quicksort

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- Fastest known **comparison-based** sorting for **arrays**
  - in the **average-case** [ $\Theta(n \log n)$ ]
  - Very widely used!
  - ... but **worst-case** is in  $\Theta(n^2)$

# Quicksort

- Like mergesort, quicksort is also a **divide and conquer** algorithm, however, they have important differences

	Dividing	Merging
Mergesort	Simple	all the work
Quicksort	all the work	Simple (Concatenation)

# Sorting Algorithms

	Worst-case	Best-case	In-place	Stable
Insertion	$O(n^2)$	$O(n)$	✓	✓
Selection	$O(n^2)$	$O(n^2)$	✓	✗
Merge	$O(n \log n)$	$O(n \log n)$	✗	✓
Quick	$O(n^2)$	$O(n \log n)$	✓	✗

# Heap Sort

# Heap Sort

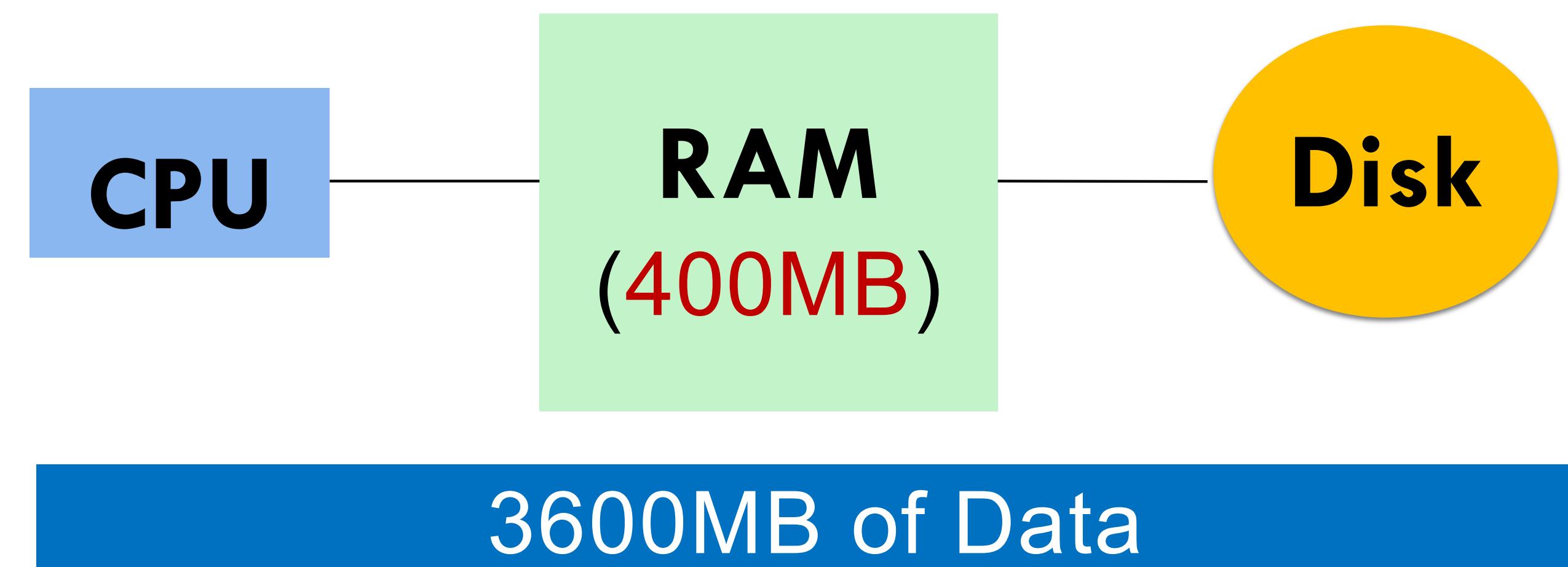
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- Sorting algorithm that uses heaps
- Given an array of unsorted numbers
  - Build a heap (Floyd's method)
  - Call removeMin( ) n times
  - We get a sorted keys!
- Total time =  $O(n \log n)$

# **External Sort**

# Sorting Big Data

- Suppose we'd like to sort **3600MB** of numbers using **400MB RAM**
- Which sorting algorithm would you use? What would you do to minimize disk accesses?



# Sorting Big Data

Suppose we'd like to sort **3600MB** of numbers using **400MB RAM**

How many reads are required?

How many MB can be sorted in one read?

How to merge and store sorted results?

# External Sort – High Level Idea

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- **Step 1: Divide** the input into smaller chunks that fit into memory.
- **Step 2: Sort** each chunk individually.
- **Step 3: Merge** the sorted chunks into a single sorted output.

# External Sorting

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- External Sorting is a technique used to sort **large amounts** of data **that cannot fit into memory** all at once.
- It involves **reading chunks of data into memory**, **sorting them**, and then **writing them back to disk**.
- Commonly used in sorting large scale datasets stored on the disk.

# External Sort

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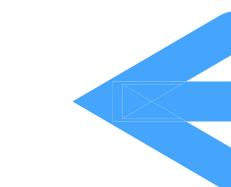
- Read 400MB of data into memory
  - Sort using a conventional method (e.g., quicksort)
  - Write sorted 400MB to a temporary file on disk
  - Repeat until all data is in sorted chunks ( $3600/400 = 9$  chunks in total)
- Read first 40MB of each sorted chunk, and merge using remaining 40MB
  - Read and write to the disk as necessary
  - Single 9-way merge is used

# Sorting Massive Data – Summary

- Need sorting algorithms that **minimize disk accesses**
- Quicksort and Heapsort both jump all over the array, leading to **expensive random disk accesses**
- Merge sort scans linearly through arrays, leading to (relatively) **efficient sequential disk access**
  - Merge sort is the basis of massive sorting
  - Merge sort can leverage multiple disks

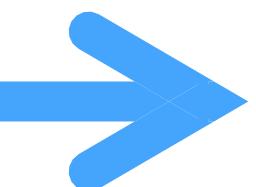
# Lower Bound for Comparison-based Sorting

$O(n^2)$



- insertion sort
- selection sort

$O(n \log n)$



- heap sort
- mergesort
- quicksort
- BST/AVL sort

Can we do better? Sort faster than  $O(n \log n)$ ?

# Lower Bound for Comparison-based Sorting

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- If the computational primitive is **comparison** (e.g.,  $<$ ,  $>$ ,  $\geq$ ,  $\leq$ ) then  $\Omega(n \log n)$  is the worst-case lower bound on sorting
- This means that **no comparison-based** sorting algorithm can ever run faster than  $\Omega(n \log n)$  on arbitrary input!
  - Proof: see book! (11.3.1)

# Question

PollEv

We are given an **Array A** of  $n$  non-negative integers in the range of 0 to  $k$ . What is the most suitable approach?

**For non-negative integers in a range from 0 to k, which of the following is most suitable method for sorting?**

Divide and conquer strategy

0%

Comparison based in-place algorithm

0%

Using an array, where each slot stores a count of a number corresponding to its index

0%

External sort

0%

None of the above

0%

## Question

We are given an **Array A** of  $n$  non-negative integers in the range of 0 to  $k$ . We wish to create another Array **C** of size  $k + 1$  such that **C[x]** is equal to the number of times  $x$  appears in the original Array **A**.

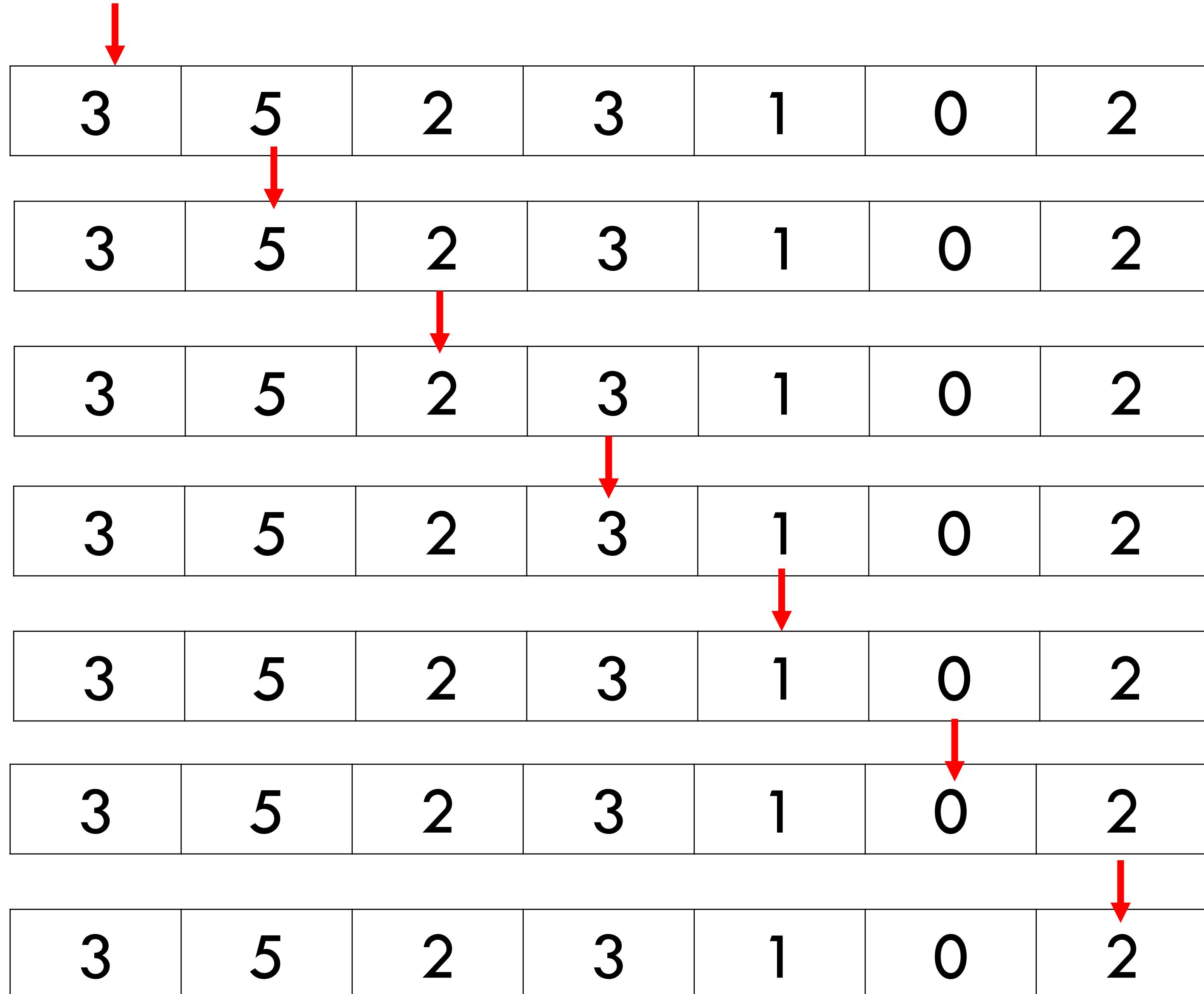
### Example:

$$A = 2, 5, 3, 0, 2, 3, 0, 3$$

$$\text{then } C = 2, 0, 2, 3, 0, 1$$

# Count Sort

# Example



0	1	2	3	4	5	6
			1			
0	1	2	3	4	5	6
			1		1	
0	1	2	3	4	5	6
		1	1		1	
0	1	2	3	4	5	6
	1	2		1		
0	1	2	3	4	5	6
1	1	2	1			
0	1	2	3	4	5	6
1	1	2	2		1	

# Sorting

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- How can we get the sorted output using C?
- What is the complexity of sorting?

# Range Search

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- How can you use the output array for range search?
- What is the complexity of this approach?

# Counting Sort – Limitation

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- Requires a **limited range of values**
- Not suitable for **large or unbounded integers**
- **Can not sort negative numbers** without modification
- Not an **in-place sorting algorithm**
- Not be used for floating numbers or string
- Inefficient for small data sets – **Wastes memory**

# Bucket Sort

# Bucket Sort

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- Non-comparative sorting algorithm that distributes elements into multiple "buckets" and sorts them individually
- Sorting numbers with a known range, especially floating-point numbers and uniformly distributed data.
- Takes  $O(n)$ , if elements are evenly distributed across buckets.

# Bucket Sort (a.k.a. Integer Sorting)

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- Uses buckets to sort integers in the range 0 to  $k - 1$
- High level Idea
  - Keep an array of  $q$  queues (or buckets)
  - Walk through the array, enqueue each key  $i$  in queue  $i$
  - Last step: Concatenate queues in order
- Works well when the keys are in a small range ( $q$  is in  $O(n)$ )
  - $n$  is the number of integers to sort

# Bucket Sort

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- How to sort numbers [0.42, 0.32, 0.23, 0.52, 0.25, 0.47, 0.51]?
- High level Idea
  - Create buckets
  - Each number is placed in the corresponding bucket
  - Sort each bucket
  - Concatenate sorted buckets

# Bucket Sort

- How to sort numbers [0.42, 0.32, 0.23, 0.25, 0.47, 0.50]?

Create buckets

Buckets	
0	0.0-0.1
1	0.1-0.2
2	0.2-0.3
3	0.3-0.4
4	0.4-0.5

Insert elements

Buckets	Elements
0	
1	
2	0.23,0.25
3	0.32
4	0.42,0.47,0.5

# Bucket Sort

- How to sort numbers [0.42, 0.32, 0.23, 0.25, 0.47, 0.50]?

Sort elements

Buckets	Elements
0	
1	
2	0.23,0.25
3	0.32
4	0.42,0.47,0.5

Concatenate Buckets

0.23,0.25,0.32,0.42,0.47,0.5

This is exactly a hash table!

# Bucket Sort – Time Complexity

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- Data is uniformly distributed –  $O(n)$
- $O(q + n)$ 
  - Linear in  $n$  but also linear in  $q$
  - time to initialize and concatenate  $q$  queues  $O(q)$
  - time to put items in queues  $O(n)$
- $O(n^2)$  when all elements are in a single bucket

# **Radix Sort**

# Radix Sort – Time Complexity

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- Sorts numbers digit by digit, starting either from the least significant digit (LSD) or from the most significant digit (MSD).
- Data is uniformly distributed –  $O(n)$
- $O(q + n)$ 
  - Linear in  $n$  but also linear in  $q$
  - time to initialize and concatenate  $q$  queues  $O(q)$
  - time to put items in queues  $O(n)$
- Worst case time complexity when working with huge numbers.  
 $O(n^2)$

# Radix Sort

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- Works well with
  - Integer
  - Fixed length string
- Limitations
  - Memory requirement
  - Data doesn't fit into buckets
  - Sparse data with outliers (when elements have huge digits)

# Sorting Algorithms

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- <https://www.toptal.com/developers/sorting-algorithms>
- <https://www.cs.usfca.edu/~galles/visualization/ComparisonSort.html>
- <https://www.youtube.com/watch?v=kPRA0W1kECg>

# Questions

