# Sorting Lower Bounds & Linear Sorting Algorithms

# **Outline for Today**

#### Sorting Lower Bounds

Comparison-based sorting algorithms

[Example] Insertion Sort, Merge Sort (revisited)

Sorting Lower Bounds

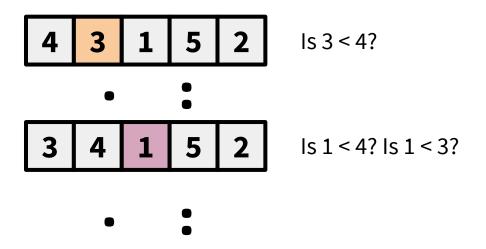
Linear-time sorting algorithms

# **Sorting Lower Bounds**

These algorithms use "comparisons" to achieve their output.

insertion\_sort and mergesort are comparison-based sorting algorithms.

A comparison compares two values. e.g. Is **A[0]** < **A[1]**? Is **A[0]** < **A[4]**? Recall, insertion sort.



mergesort: comparison happens in the merge subroutine. (explain on board) select\_k is a comparison-based algorithm (compare each value with pivot)

Next week, we'll learn about a randomized comparison-based sorting algorithm called quicksort.

**Theorem:** Any deterministic comparison-based sorting algorithm requires  $\Omega(n \log(n))$ -time.

Remember: not all sorting algorithms require  $\Omega(n \log(n))$  time, some algorithms can be faster than this.

#### Keywords:

Deterministic -> the list will be accurately sorted for sure when the algorithm terminates. There are some algorithms sort the list accurately only with a probability, or sort the list approximately, but are faster.

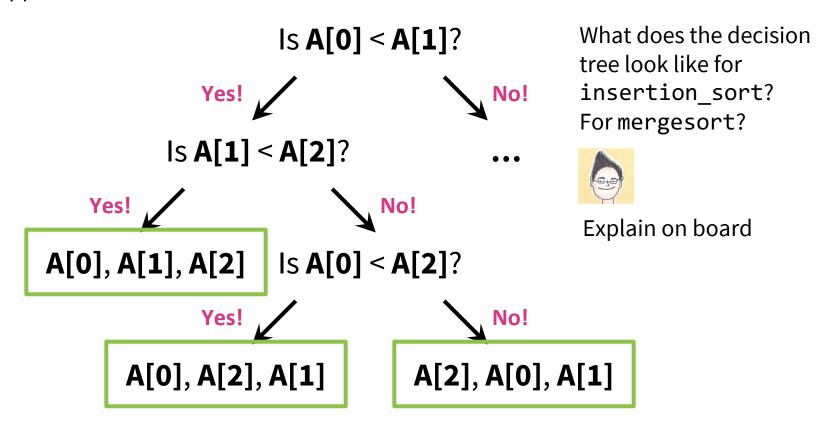
Comparison-based -> there are some algorithms do not need to do comparison for sorting, e.g. counting sort (will discuss it later)

#### **Proof:**

Hmm ...

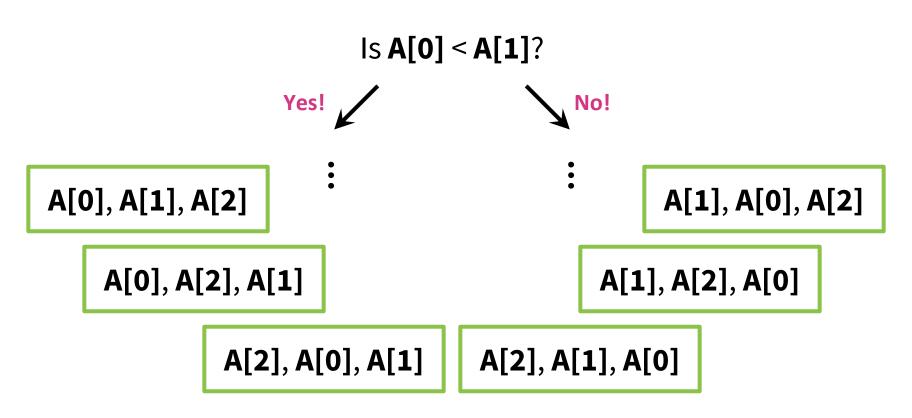
We can represent the comparisons made by a comparisonbased sorting algorithm as a decision tree.

Suppose we want to sort three items in **A**.



The leaves are all of the possible orderings of the items.

The worst-case runtime must be at least  $\Omega$  (length of the longest path).



#### How long is the longest path?

At least how many leaves must this decision tree have?

What is the depth of the shallowest tree with this many leaves?

How long is the longest path?

At least how many leaves must this decision tree have? n!

What is the depth of the shallowest tree with this many leaves? log(n!)

The longest path is at least log(n!), so the worst-case runtime must be at least  $\Omega(log(n!)) = \Omega(n log(n))$ .

Explain on board: the balanced binary tree gives the shortest depth.

The Stirling's approximation:  $n! \sim \sqrt{2\pi n} \left(\frac{n}{e}\right)^n$ Explain on board:  $\Omega(\log(n!)) = \Omega(n \log(n))$ 

**Theorem:** Any deterministic comparison-based sorting algorithm requires  $\Omega(n \log(n))$ -time.

#### **Proof:**

Any deterministic comparison-based sorting algorithm can be represented as a decision tree with n! Leaves.

The worst-case runtime is at least the depth of the decision tree.

All decision trees with n! leaves have depth  $\Omega(n \log(n))$ .

Therefore, any deterministic comparison-based sorting algorithm requires  $\Omega(n \log(n))$ -time

### **Beyond Comparisons**

But then what's this nonsense about linear-time sorting algorithms?

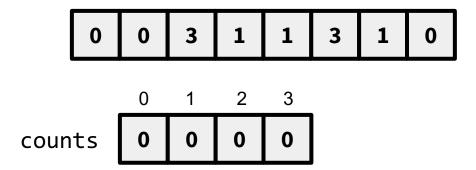
We achieve O(n) worst-runtime if we make assumptions on the input. e.g. They are integers that range from 0 to k-1.

```
algorithm counting_sort(A, k):
    # A consists of n ints, ranging from
    # 0 to k-1
    counts = [0 * k] # list of k zeros
    for a_i in A:
        counts[a_i] += 1
    result = []
    for a_i = 0 to length(counts)-1:
        append counts[a_i] a_i's to results
    return results
```

Runtime: O(n+k)

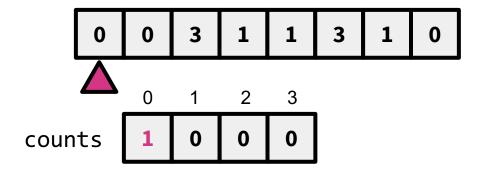
Suppose A consists of 8 ints ranging from 0 to 3.

counting\_sort(A, 4)



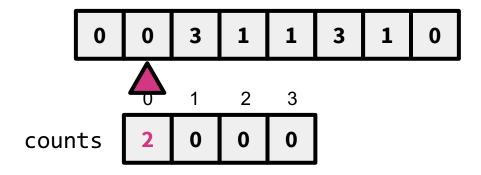
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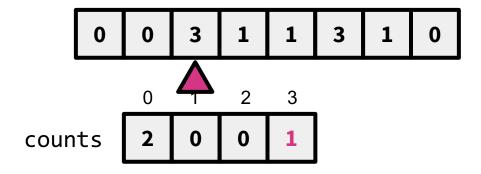


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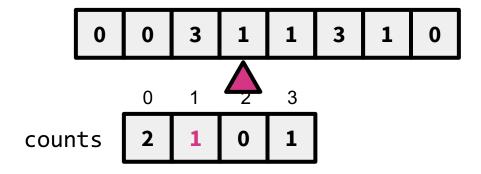
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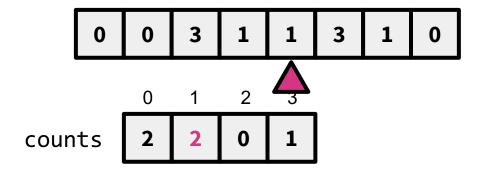
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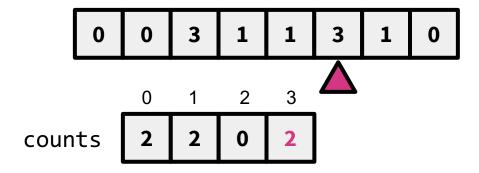


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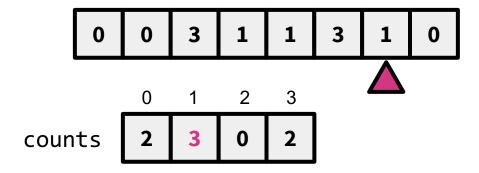
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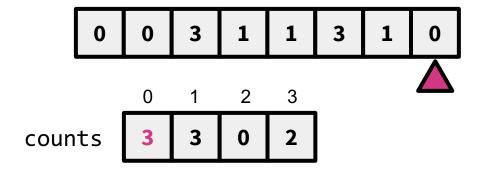


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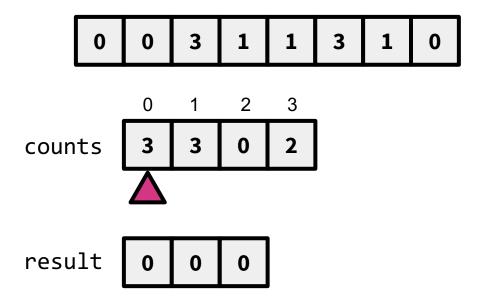


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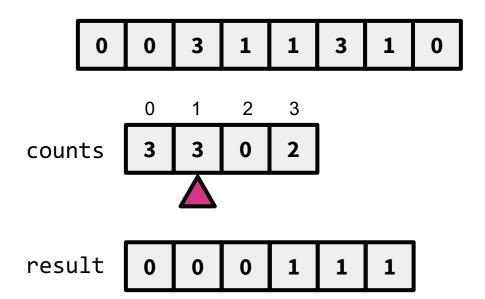
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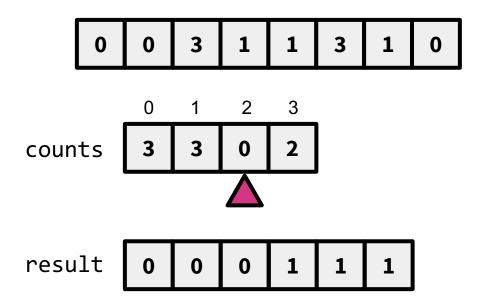
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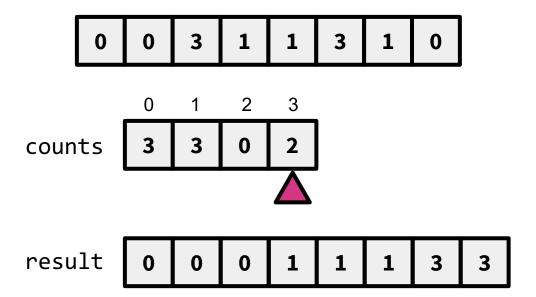
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Runtime: O(n+k)

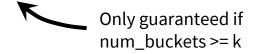
#### **5-Minute Break**

# **Bucket Sort**

#### **Bucket sort**

```
algorithm bucket_sort(A, k, num_buckets):
    # A consists of n (key, value) pairs,
    # with keys ranging from 0 to k-1
    buckets = [[] * num_buckets]
    for key, value in A:
        buckets[get_bucket(key)].append((key, value))
    if num_buckets < k:
        for bucket in buckets:
            stable_sort(bucket) by their keys
    result = concatenate buckets by their values
    return result</pre>
```

Runtime: O(n+k) or O(nlogn)



#### **Bucket sort**

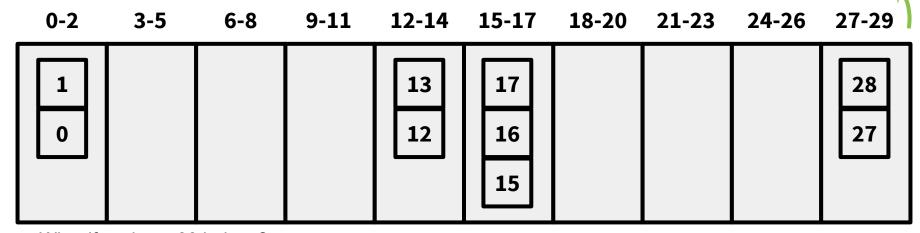
Two cases for k and num\_buckets in bucket\_sort:

- (1) k ≤ num\_buckets: At most one key per bucket, so buckets do not require an additional stable\_sort to be sorted (similar to counting\_sort).
- (2) k > num\_buckets: Maybe multiple keys per bucket, so buckets require an additional stable\_sort to be sorted.

Suppose k = 30 and num\_buckets = 10. Then we group keys 0 to 2 in the same bucket, 3 to 5 in the same bucket, etc.

A= [17, 13, 16, 12, 15, 1, 28, 0, 27] produces:

Only the keys in the (key, value) pairs are shown here, and all of the buckets require stable\_sort.



What if we have 30 bukets?

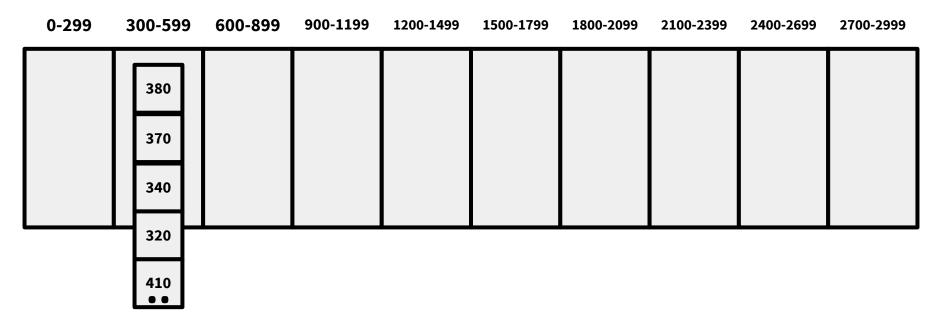
# Bucket sort, case (2)

#### Why O(nlogn) in case (2)?

With multiple keys per bucket, a bucket might receive all of the inserted keys.

Suppose the bucket\_sort caller specifies k = 3000 and num\_buckets = 10, but then inserts elements all from the same bucket.

A = [380, 370, 340, 320, 410, ...] would need to stable\_sort all of the elements in the original list since they all fall in the same bucket.



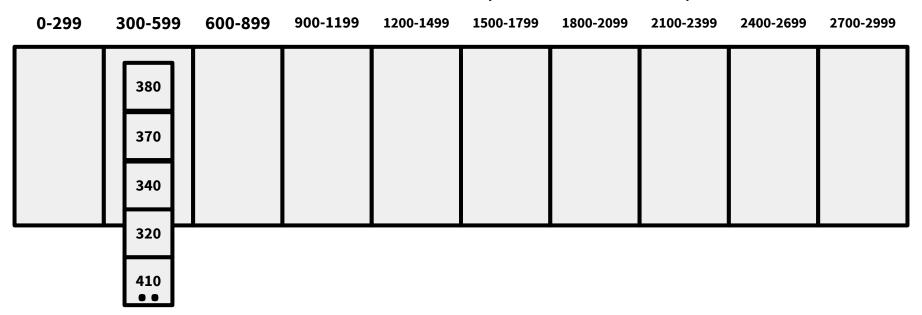
# Bucket sort, case (2)

#### What to do in practice?

Find the exact smallest and largest number in the list (costs O(n)), then design more tight buckets to split numbers into the buckets as equally as possible.

Explain on board: the time when equally splitting the numbers into b buckets.

Note: Stable sort means the order of two equal numbers are kept as before.



```
algorithm radix_sort(A, d, k):
    # A consists of n d-digit ints, with
    # digits ranging 0 -> k-1
    for j = 0 to d-1:
        A_j = A converted to (key, value) pairs, where
             key is the jth digit of value
        result = bucket_sort(A_j, k, k)
        A = result
    return A
```

Runtime: O(d(n+k))

Suppose **A** consists of 8 3-digit ints, with digits ranging from 0 to 9.

radix\_sort(A, 3, 10)

Α

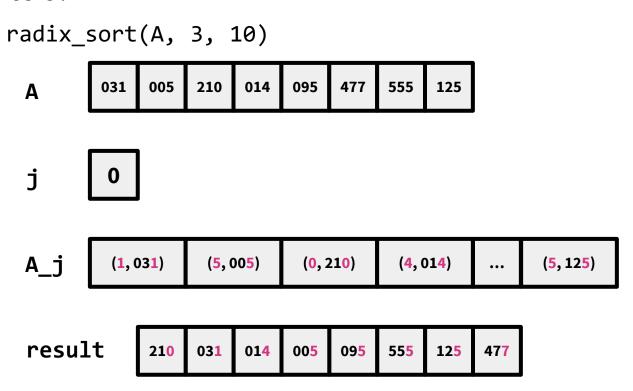


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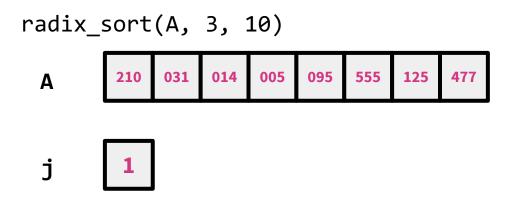
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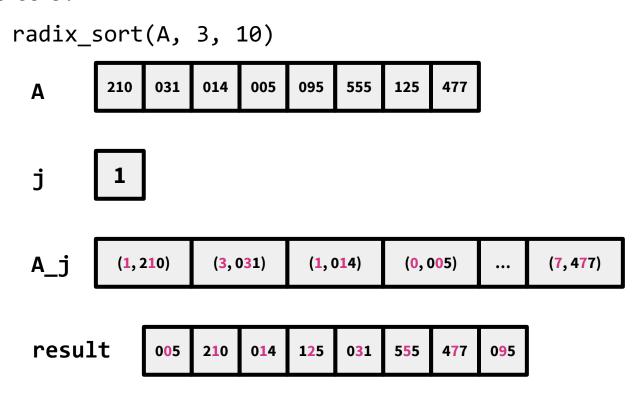
A 031 005 210 014 095 477 555 125

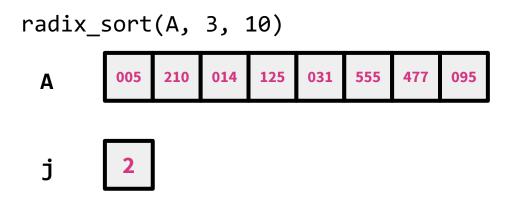
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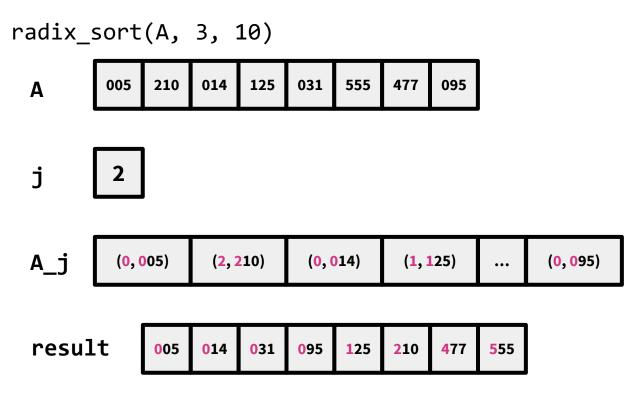


Explain on board: using bucket sort to sort A j with 10 buckets (bucket 0 to bucket 9)









**Lemma:** If **A** is sorted by its x least-significant digits by the end of iteration j = x of the loop, then **A** will be sorted by its x+1 least-significant digits by the end of iteration j = x+1 of the loop.

#### **Proof:**

Since bucket\_sort is stable, the elements within each bucket are still sorted by their x least-significant digits.

(E.g., in the second round 210 and 014 are still sorted on 0 and 4, although the middle digit are both 1.)

bucket\_sort sorts **A** by the x+1 digit of the elements, so the elements are sorted by their x+1 least-significant digits.

**Theorem:** Radix sort sorts the input list.

#### **Proof:**

At by the end of the 0-th iteration of the loop, **A** is sorted by its 0-th least-significant digits.

By our lemma, if **A** is sorted by its x least-significant digits by the end of iteration j = x of the loop, then **A** will be sorted by its x+1 least-significant digits by the end of iteration j = x+1 of the loop.

The loop terminates at the start of iteration j = d. The collection of d-digit integers in **A** are sorted by their d least-significant digits, which implies that **A** is sorted when the loop ends.

### Summary

#### **Sorting lower bounds**

For any deterministic comparison-based sorting algorithm, the lower bound of computing time is  $\Omega(n \log(n))$ .

#### **Linear Sorting Algorithms**

If we know extra information about the input list, we may design linear-time sorting algorithms.

**Counting Sort** 

**Bucket Sort** 

Radix Sort

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For any deterministic comparison-based sorting algorithm, the lower bound of computing time is  $\Omega(n \log(n))$ .

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If we know extra information about the input list, we may design linear-time sorting algorithms.

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**Radix Sort** 

Acknowledgement: Part of the materials are adapted from Mary Wootter, Virginia Williams and David Eng's lectures on algorithms. We appreciate their contributions.