

CSE 332 Summer 2026

Lecture 11: Sorting 2

Michael Whitmeyer

<http://www.cs.uw.edu/332>

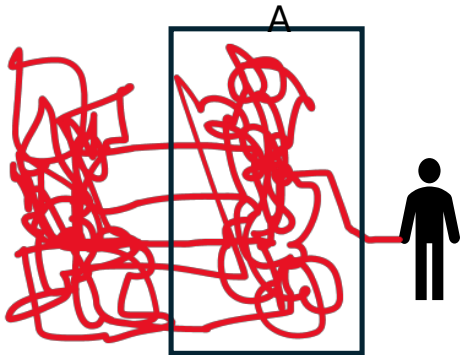
Properties To Consider

- Worst case running time
- In place:
 - We only need to use the pre-existing array to do sorting
 - Constant extra space (only some additional variables needed)
- Adaptive
 - The running improves as the given list is closer to being sorted
 - It should be linear time for a pre-sorted list, and nearly linear time if the list is nearly sorted
- Online
 - We can start sorting before we have the entire list.
- Stable
 - “Tied” elements keep their original order

Divide And Conquer Sorting

- Divide and Conquer:
 - Recursive algorithm design technique
 - Solve a large problem by breaking it up into smaller versions of the same problem

Divide and Conquer



- **Base Case:**

- If the problem is “small” then solve directly and return

- **Divide:**

- Break the problem into subproblem(s), each smaller instances

- **Conquer:**

- Solve subproblem(s) recursively

- **Combine:**

- Use solutions to subproblems to solve original problem

Merge Sort Pseudocode

mergesort(L):

ms_helper(L, 0, L.length())

mshelper(L, low, high):

IF (low == high) RETURN // Base Case

L is length 1

mid = (low+high)/2

ms_helper(low, mid)

ms_helper(mid+1, high)

merge(L, low, mid, high)

Merge Sort Properties

- Worst case running time

- $\Theta(n \log n)$

- In place:

- NO!

- We need a second array to merge into

- Adaptive

- NO!

- Running time is the same regardless of original list's order

- Online

- NO!

- We need all elements before we can divide

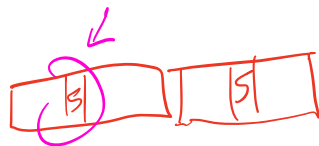
- Stable

- YES!

- When merging, and there's a tie, choose the element from the left

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

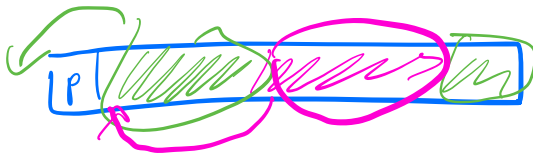
merging
↓



Quicksort Vs. Mergesort

- Like Mergesort:
 - Divide and conquer
 - $O(n \log n)$ run time (kind of...)
- Unlike Mergesort:
 - Divide step is the “hard” part
 - *Typically* faster than Mergesort

Quicksort Overview



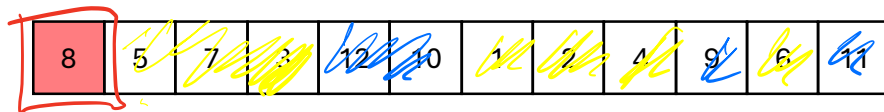
Idea: pick a **pivot** element, recursively sort two sublists around that element

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

Partition (Divide step)

Given: a list, a pivot p

Start: unordered list



Goal: All elements $< p$ on left, all $> p$ on right

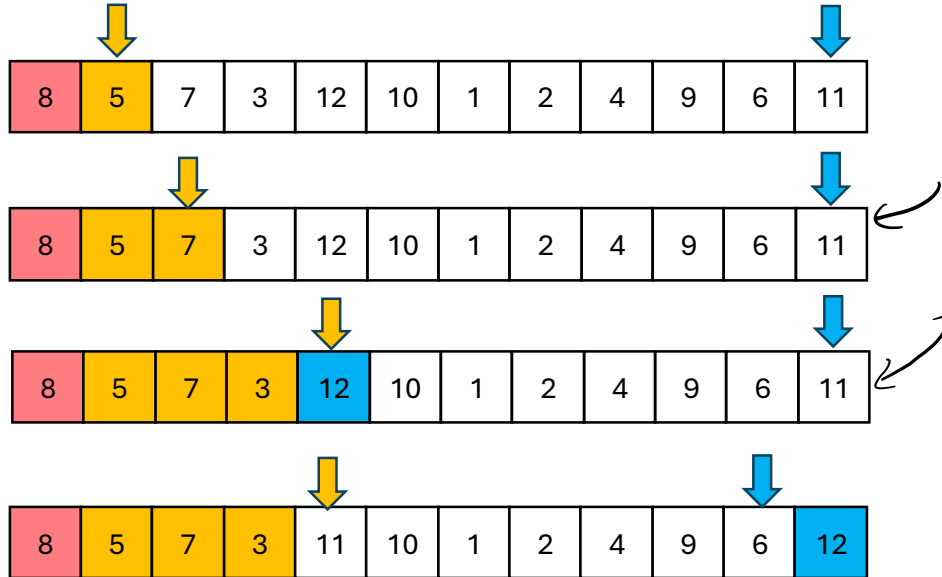


Partition Procedure (Slide 1 of 2)

If **Begin** value $< p$, move **Begin** right

Else swap **Begin** value with **End** value, move **End** Left

Done when **Begin** = **End**

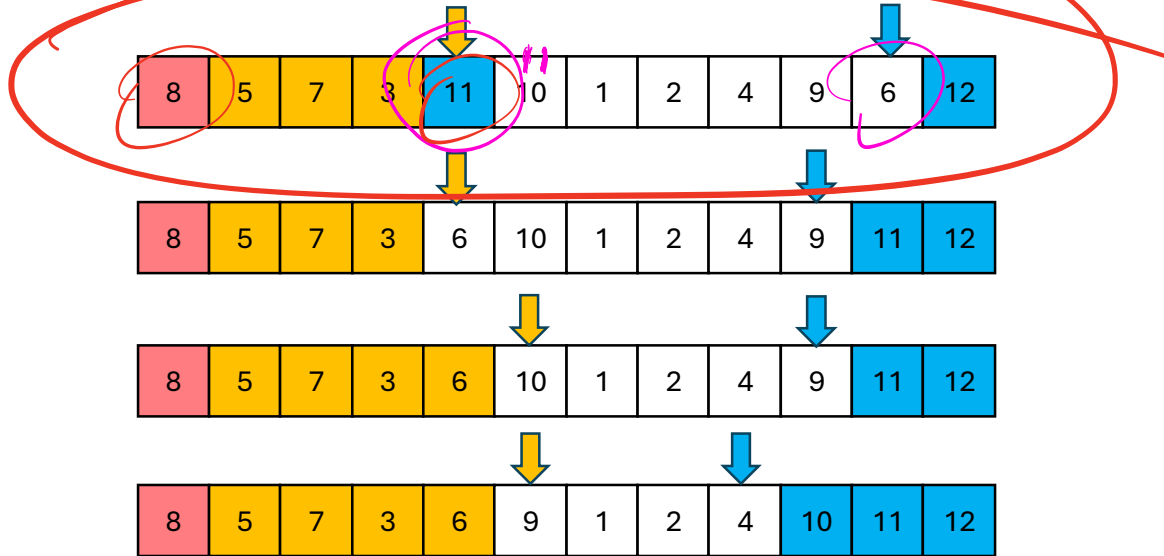


Partition Procedure (Slide 2 of 2)

IF **Begin** value < p , move **Begin** right

ELSE swap **Begin** value with **End** value, move **End** Left

Done when **Begin** = **End**

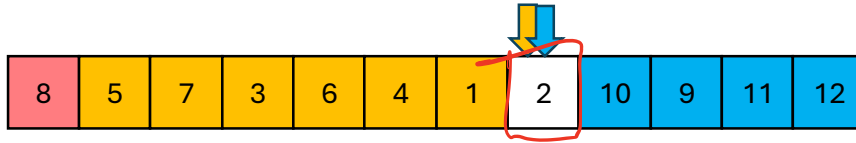


Partition – Begin Meets End (Case 1)

If **Begin** value $< p$, move **Begin** right

Else swap **Begin** value with **End** value, move **End** Left

Done when **Begin** = **End**



Case 1: meet at element $< p$

Swap p with **pointer position** (2 in this case)

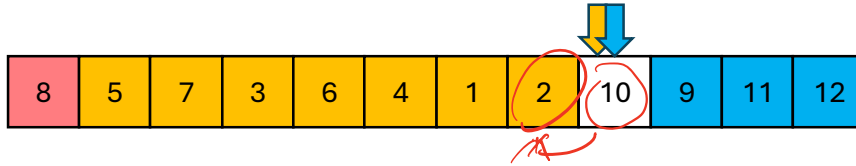


Partition – Begin Meets End (Case 2)

If **Begin** value $< p$, move **Begin** right

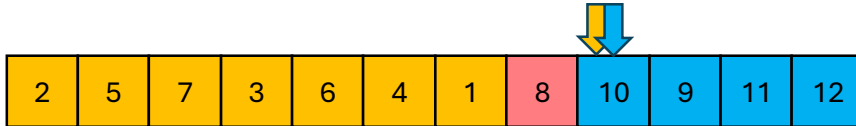
Else swap **Begin** value with **End** value, move **End** Left

Done when **Begin** = **End**



Case 2: meet at element $> p$

Swap p with **value to the left** (2 in this case)



Partition Summary

1. Put p at beginning of list
2. Put a pointer (Begin) just after p , and a pointer (End) at the end of the list
3. While $\text{Begin} < \text{End}$:
 1. If Begin value $< p$, move Begin right
 2. Else swap Begin value with End value, move End Left
4. If pointers meet at element $< p$: Swap p with pointer position
5. Else If pointers meet at element $> p$: Swap p with value to the left

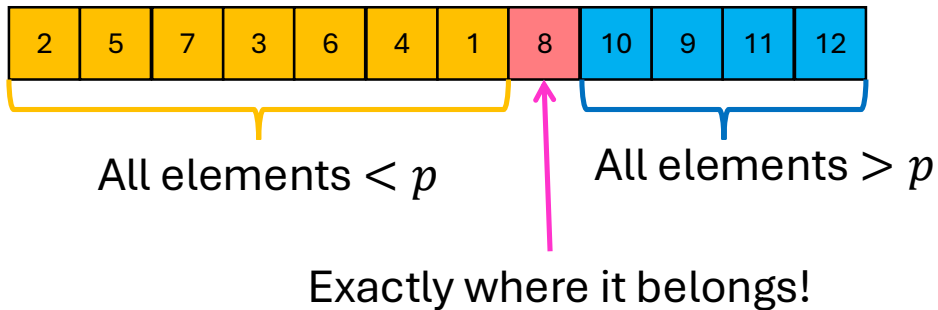
Run time? $O(n)$

Partition Summary

1. Put p at beginning of list
2. Put a pointer ($Begin$) just after p , and a pointer (End) at the end of the list
3. While $Begin < End$:
 1. If $Begin$ value $< p$, move $Begin$ right
 2. Else swap $Begin$ value with End value, move End Left
4. If pointers meet at element $< p$: Swap p with pointer position
5. Else If pointers meet at element $> p$: Swap p with value to the left

Run time? $O(n)$

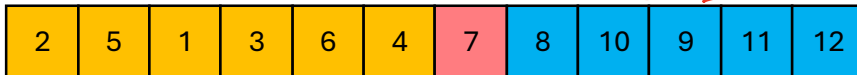
Conquer



Recursively Quicksort™ Left and Right sublists

Quicksort Run Time (Best)

If the **pivot** is always the median:

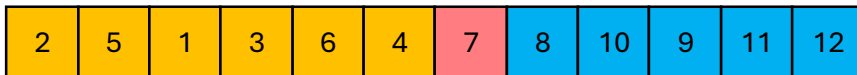


Then we divide in half each time

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

Quicksort Run Time (Best)

If the **pivot** is always the median:



Then we divide in half each time

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

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

Quicksort Run Time (Worst)

If the pivot is always at the extreme:

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

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

Then we shorten by 1 each time

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

$$+ n-1 + T(n-2)$$

$$+ n-2 + T(n-3)$$

$$n + n-1 + n-2 + \dots + 1$$

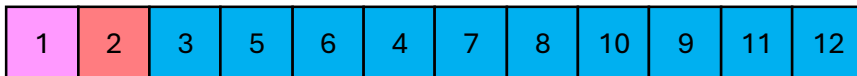
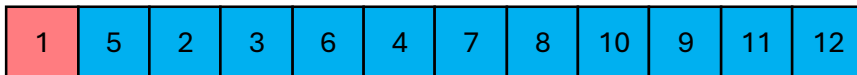
$$= \sum_{i=1}^n i$$

$$= \frac{n(n+1)}{2}$$

$$T(n) = n + T(n-1) = O(n^2)$$

Quicksort Run Time (Worst)

If the pivot is always at the extreme:

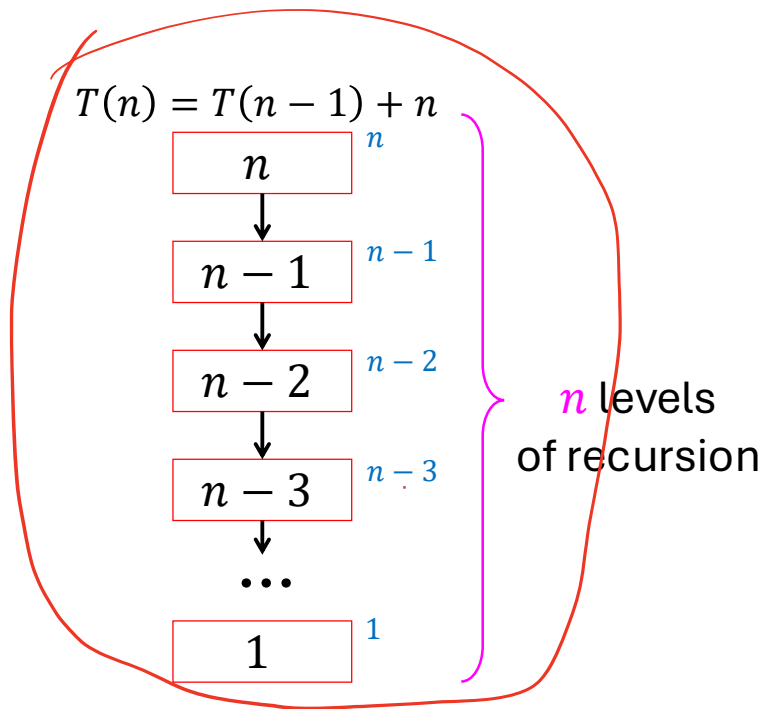


Then we shorten by 1 each time

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

$$T(n) = O(n^2)$$

Quicksort Worst Case Tree Method



$$T(n) = \sum_{i=0}^{n-1} n - i = \sum_{i=1}^n i = \frac{n(n+1)}{2}$$
$$= \Theta(n^2)$$

Good Pivot

- What makes a good Pivot?
 - Roughly even split between left and right
 - Ideally: median
- There are ways to find the median in linear time, but it's complicated and slow and you're better off using mergesort
- In Practice:
 - Pick a random value as a pivot
 - Pick the middle of 3 random values as the pivot

$O(n)$

421

$$T(n) = n + T\left(\frac{n}{4}\right) + T\left(\frac{3n}{4}\right)$$

Quick Sort Properties

- Worst case running time? $O(n^2)$
- In place? *controversial*
- Adaptive? *no!*
- Online? *no*
- Stable? *no.*

Quick Sort Properties

- Worst case running time
 - $\Theta(n^2)$, but “almost always” $\Theta(n \log n)$ in practice
 - Better constants than ~~merge sort~~
- In place:
 - Kinda...
 - We swap within the given array, but do so using recursion, so we need space for the stack frames
 - Different textbooks disagree on whether call-stack space “counts”
- Adaptive
 - NO!
- Online
 - NO!
 - We need all elements before we can divide
- Stable
 - NO!
 - Partition procedure may rearrange tied elements

$$O(n \log n)$$

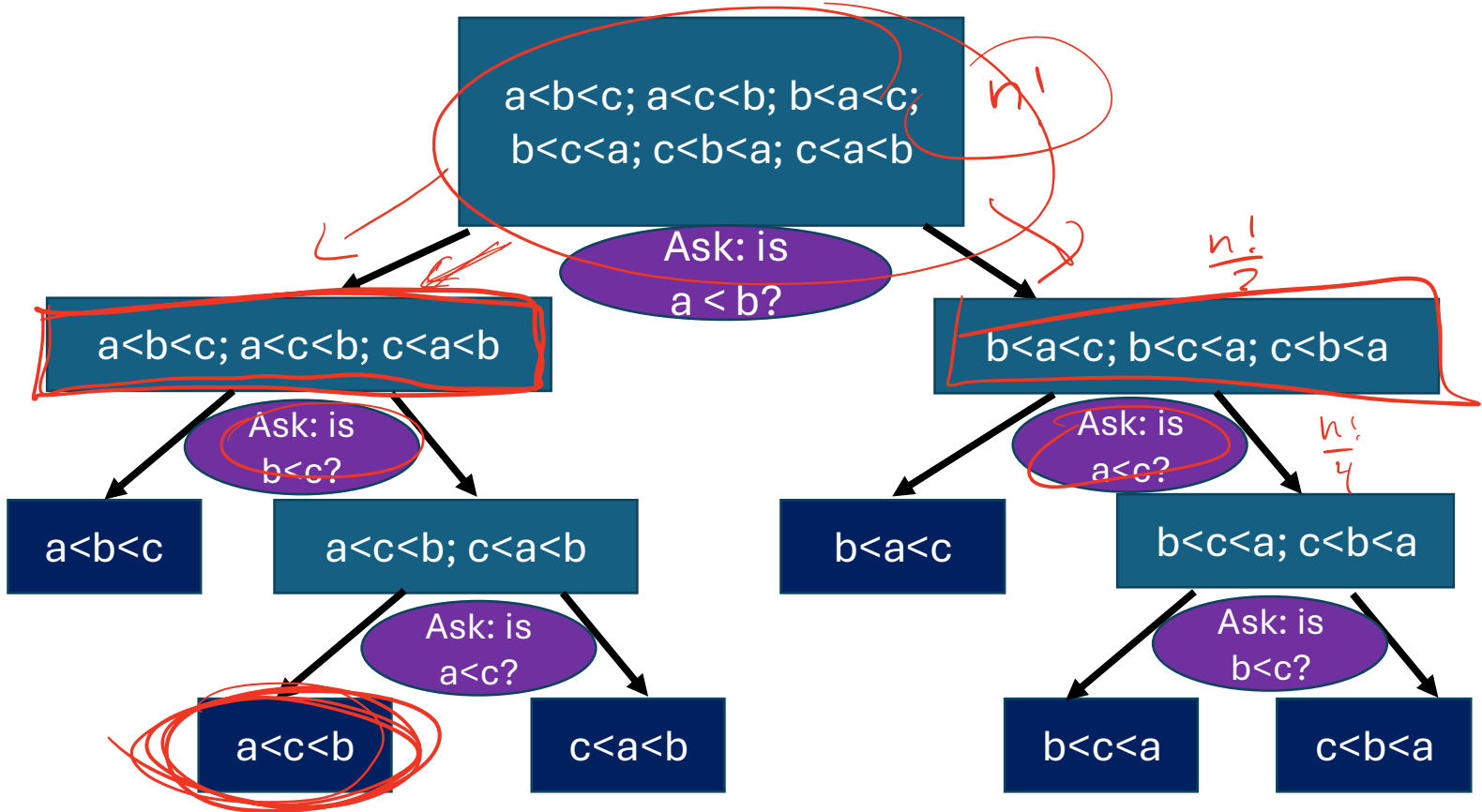
- We keep seeing a worst case $O(n \log n)$ runtime
 - Merge, heap, quick (with a good pivot)
- Can we do better?

Theorem: $\Omega(n \log n)$

- We keep seeing a worst case $O(n \log n)$ runtime
 - Merge, heap, quick (with a good pivot)
- Can we do better?
- **Theorem:** Any comparison-based Sorting algorithm requires $\Omega(n \log n)$ comparisons

Proving the Lower Bound

- **Theorem:** Any comparison-based sorting algorithm requires $\Omega(n \log n)$ comparisons
- Proof uses “**decision trees**”
 - A tree showing the decisions our code will make
 - Nodes of tree labelled with “ $a < b ?$ ”



Proving the Lower Bound

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

- **Proof:**

- ~~Worst case runtime~~ = ^{# of comparisons} height/depth of tree
- # leaves = # permutations = $n!$
- Depth lower bound =

Key observation: at every ^{internal} node, ≥ 1 child with $\geq \frac{1}{2}$ of the remaining possible permutations.

$$\log_2(n!) \geq \log_2\left(\left(\frac{n}{2}\right)^{\frac{n}{2}}\right) \geq \frac{n}{2} \cdot \log_2\left(\frac{n}{2}\right) = \Omega(n \log n)$$

Proving the Lower Bound

- **Theorem:** Any comparison-based sorting algorithm requires $\Omega(n \log n)$ comparisons
- **Proof:**
 - Worst case runtime = height of tree
 - # leaves = # permutations = $n!$
 - Depth lower bound = $\log_2(n!)$

Calculation: $\log_2(n!) = \log_2(n) + \log_2(n-1) + \log_2(n-2) + \dots + \log_2(1)$

$$\geq \log_2\left(\frac{n}{2}\right) + \log_2\left(\frac{n}{2}\right) + \dots + \log_2\left(\frac{n}{2}\right) \text{ (only } n/2 \text{ copies)}$$

$$\geq \frac{n}{2} \log_2\left(\frac{n}{2}\right) = n/2(\log_2(n) - 1) = \Omega(n \log n)$$

“Linear Time” Sorting Algorithms

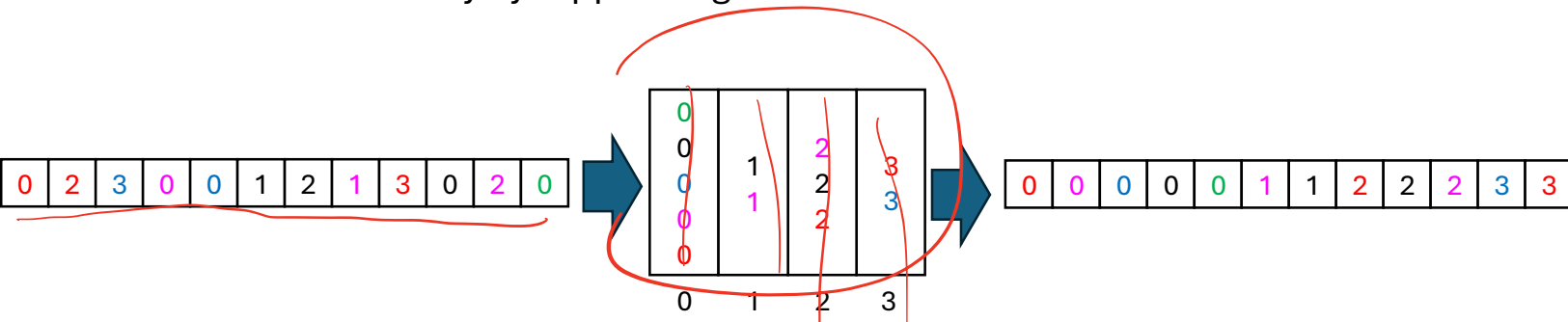
integers are small.

- Useable when you are able to make additional assumptions about the contents of your list (beyond the ability to compare)
 - Examples:
 - The list contains only positive integers less than k
 - The number of distinct values in the list is much smaller than the length of the list
- The running time expression will always have a term other than the list's length to account for this assumption
 - Examples:
 - Running time might be $\Theta(k \cdot n)$ where k is the range/count of values

sleep sort: make timers for each integer, when timers go off, add to array.
 $O(n)$

BucketSort

- Assumes the array contains integers between 0 and $k - 1$ (or some other small range)
- Idea:
 - Use each value as an index into an array of size k
 - Add the item into the “bucket” at that index (e.g. linked list)
 - Get sorted array by “appending” all the buckets



BucketSort Running Time

- Create array of k buckets
 - Either $\Theta(k)$ or $\Theta(1)$ depending on some things...
- Insert all n things into buckets
 - $\Theta(n)$
- Empty buckets into an array
 - $\Theta(n + k)$
- Overall:
 - $\Theta(n + k)$
- When is this better than mergesort?

Properties of BucketSort

- In-Place?
- Adaptive?
- Stable?

Properties of BucketSort

- In-Place?
 - No
- Adaptive?
 - No
- Stable?
 - Yes!

RadixSort

- Radix: The base of a number system
 - We'll use base 10, most implementations will use larger bases
- Idea:
 - BucketSort by each digit, one at a time, from least significant to most significant

10	80	40	32	25	82	99	10	11	90	55	51	24	80	01	12
3	1	1	3	5	3	9	1	3	1	5	2	5	0	8	1
0	1	2	3	4	5	6	7	8	9	10	11	12	13	14	15

Place each element
into a “bucket”
according to its **1's**
place



800	801 401 101 901 121	512	103 323 823 113		255 555 245			018	999
0	1	2	3	4	5	6	7	8	9

RadixSort – Re-Bucket by 10s

- Radix: The base of a number system
 - We'll use base 10, most implementations will use larger bases
- Idea:
 - BucketSort by each digit, one at a time, from least significant to most significant

800	801 401 101 901 121	512	103 323 823 113		255 555 245			018	999
0	1	2	3	4	5	6	7	8	9

Place each element
into a “bucket”
according to its **10's**
place



800 801 401 101 901 103	512 113 018	121 323 823		245	255 555				999
0	1	2	3	4	5	6	7	8	9

RadixSort – Re-Bucket by 100s

- Radix: The base of a number system
 - We'll use base 10, most implementations will use larger bases
- Idea:
 - BucketSort by each digit, one at a time, from least significant to most significant

800									
801									
401	512	121			255				
101	113	323		245	555				999
901	018	823							
103									
0	1	2	3	4	5	6	7	8	9

Place each element
into a “bucket”
according to its **100’s**
place



	101							800	901
018	103	245	323	401	512			801	999
	113	255			555			823	
	121								
0	1	2	3	4	5	6	7	8	9

RadixSort – Empty Buckets into Output Array

- Radix: The base of a number system
 - We'll use base 10, most implementations will use larger bases
- Idea:
 - BucketSort by each digit, one at a time, from least significant to most significant

018	101 103 113 121	245 255	323	401	512 555			800 801 823	901 999
0	1	2	3	4	5	6	7	8	9



Convert back into an array

018	101	103	113	121	245	255	323	401	512	555	800	801	823	901	999
0	1	2	3	4	5	6	7	8	9	10	11	12	13	14	15

RadixSort Running Time

- Suppose largest value is m
- Choose a radix (base of representation) b
- BucketSort all n things using b buckets
 - $\Theta(n + b)$
- Repeat once per each digit
 - $\log_b m$ iterations
- Overall:
 - $\Theta(n \log_b m + b \log_b m)$
- In practice, you can select the value of b to optimize running time
- When is this better than mergesort?