

Sorting Algorithms

- The act of arranging data in groups according to some property of the data
- Sorting is a well-studied problem in CS
 - Many algorithms can be made faster if you first sort the data you operate on
- Examples in these slides will sort `ints`, but will work on other data types (unless indicated otherwise)

Determining the Sort Order

- Values are sorted according to a “natural” ordering
 - Numerical values are sorted in ascending order
 - String sorted lexicographically
 - Objects in a class that implements the `Comparable` interface are sorted using the `compareTo` method

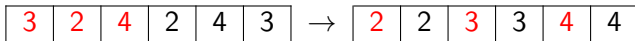
The Comparable Interface

- Has a single method: `compareTo`
 - The calling object is compared to the argument
 - Returns:
 - a negative integer if the calling object is less than the argument
 - 0 if the calling object is equal to the argument
 - a positive integer if the calling object is greater than the argument
 - Defines the “natural” ordering for objects in the class

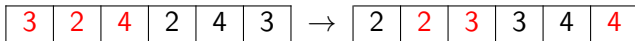
Stable Sorts

- **Stable Sort**

- A sort in which the relative order of equal elements is preserved
- Suppose you want to sort a deck of cards first by suit, then by value
 - A stable sort will ensure that the order of the suits will remain the same within the same value



Stable Sort



Non-Stable Sort

Sorting Performance

- Most single threaded sorts fall into one of two complexity categories
 - $O(n^2)$
 - $O(n \lg(n))$
- All following examples assume an array named `arr` with length `n`
 - All examples will sort in ascending order

Selection Sort

- One of the simplest sorts
- Find the smallest value in the range 0 to $n-1$ and put it in index 0 (swap the value at index 0 with the value at the found index)
- Repeat for index 1 , index 2 , ...
 - Essentially, select the element that should go at an index and put it there

Selection Sort Pseudocode

```
def find_min_index(arr, start):  
    small_idx = start  
    for (int i=start; i<arr.len; i++):  
        if arr[i] < arr[small_idx]:  
            small_idx = i  
    return small_idx  
  
def selection_sort(arr):  
    for (int idx=0; i<arr.len; idx++):  
        small_idx = find_min_index(arr, idx)  
        swap(arr[idx], arr[small_idx])
```


Selection Sort Example

- Legend:

↑↑ idx

↑ small_idx

- Original Array

6 2 0 1 7

- Pass 1: idx == 0

6 2 0 1 7

↑↑ ↑

0 2 6 1 7

↑↑ ↑

- Pass 2: idx == 1

0 2 6 1 7

↑↑ ↑

0 1 6 2 7

↑↑ ↑

- Pass 3: idx == 2

0 1 6 2 7

↑↑ ↑

0 1 2 6 7

↑↑ ↑

Selection Sort Example

- Pass 4: $idx == 3$

0 1 2 6 7

↑↑

↑

- Pass 5: $idx == 4$

0 1 2 6 7

↑↑

↑

- Selection sort is not stable
 - Data gets additionally shuffled as the sort occurs
- Selection sort has very consistent speed
 - It's not very fast, but it will take roughly the same amount of time for a given array size

Selection Sort Complexity

- Call the length of the array n
- `find_min_index(arr, start)` takes, worst case, n times through the loop
- `small_idx = find_min_index(arr, i)` runs n times
- `swap`, while a function, is fast (4 operations ish)
 - Thus: Selection Sort is $O(n^2)$
 - As written (using indexing operations), this analysis assumes $O(1)$ random access; if using a `LinkedList`, `find_min_index` would be $O(n^2)$...

Insertion Sort

- Insertion sort works on the following principle:
 - A list of one element is always sorted with itself
 - Assume elements at indexes 0 through i are in sorted order
 - Take the element at index $i+1$ and insert it into its proper place
 - Swap the element back through the list as long as it is smaller than its predecessor
 - Start i at 1, and increment i until the entire list is sorted!

Insertion Sort Pseudocode

```
def insertion_sort(arr):  
    for(i = 1; i < arr.len; i++):  
        j = i  
        while (j > 0 && arr[j] < arr[j-1])  
            swap(arr[j], arr[j-1])
```

Insertion Sort Example

- Legend:

$\uparrow \uparrow i$
 $\uparrow j-1$

- Original Array

6 2 0 1 7

- Pass 1: $i == 1$

6 2 0 1 7

$\uparrow \uparrow$

2 6 0 1 7

$\uparrow \uparrow$

- Pass 2: $i == 2$

2 6 0 1 7

$\uparrow \uparrow$

2 0 6 1 7

$\uparrow \uparrow$

0 2 6 1 7

$\uparrow \uparrow$

Insertion Sort Example

- Pass 3: $i == 3$

0 2 6 1 7

 ↑ ↑

0 2 1 6 7

 ↑ ↑

0 1 2 6 7

 ↑ ↑

- Pass 4: $i == 4$

0 1 2 6 7

 ↑ ↑

- Insertion sort, when implemented as above, is a stable sort
- Potentially improves on selection sort's performance by not sorting parts of the list that are already sorted

Insertion Sort Complexity

- To determine the complexity of insertion sort, we must count the number of times the inner loop runs
- When $i==1$, the inner loop runs at most once
- When $i==2$, the inner loop runs at most twice
- ...
- The inner loop runs a total of $1 + 2 + \dots + (n - 1) + n$ times (roughly)
 - In summation notation: $\sum_{i=1}^n i = \frac{n(n+1)}{2} = \frac{n^2+n}{2}$
 - This reduces to $O(n^2)$

Comparing Insertion and Selection

Selection Sort

- Not stable
- Worst, average, and best time: $O(n^2)$
- In place sort (no new copies of data)

Insertion Sort

- Stable
- Worst time: $O(n^2)$
- Average time: $O(n^2)$
- Best time: $O(n)$ (list already sorted)
- Good for:
 - Short lists
 - Nearly sorted lists
- In place sort (no new copies of data)

Faster Sorting

- Selection and insertion sort are *comparison-based* sorting algorithms
 - Items are sorted by comparing elements to each other
- For any comparison-based sorting algorithm, it can be rigorously shown that:
 - The worst time is *at least* $O(n \lg(n))$
 - The average time is *at least* $O(n \lg(n))$
- Merge Sort and Quick Sort both have an average time of $O(n \lg(n))$
 - linear-logarithmic
 - Both selection and insertion sort are quadratic ($O(n^2)$)

Merge Sort

- Use *Divide and Conquer* strategies to solve the problem
- Note that with selection and insertion sorts, halving the problem size quarters the runtime, so...
- Divide the array in half, and recursively apply merge sort to each half
 - There are variations on this
 - Some implementations continue until the list size is < 7 , then use insertion sort
 - Some implementations keep splitting until the list size is 1, which is inherently sorted
- Once each half is sorted, merge the sorted arrays

Two Way Merge

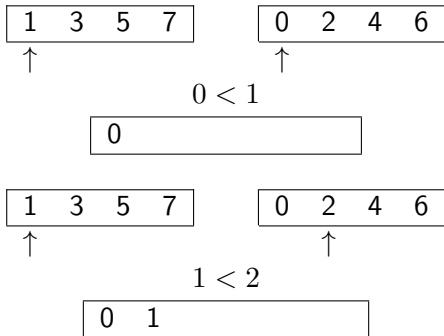
- Given two sorted lists, combine them into a single sorted list
- Continually take the smallest element from each list and add it to the new list
- Best and worst case is $O(n)$

Merge Example

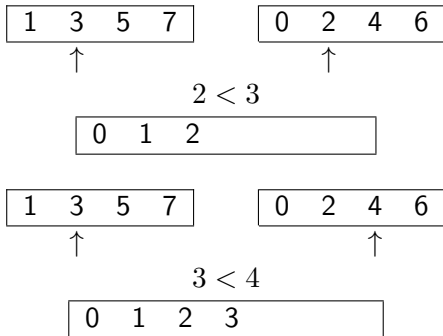
Merge the following arrays:

1 3 5 7 0 2 4 6

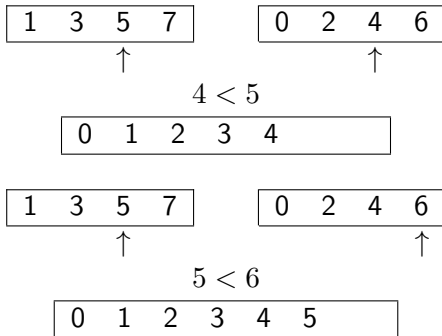
Merge Example



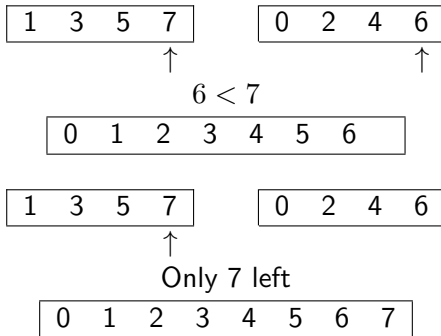
Merge Example



Merge Example



Merge Example



Merge Pseudocode

```
def merge(A, B):  
    int idxA = 0, idxB = 0  
    merged = list()  
    while (idxA < A.len && idxB < B.len):  
        if (A[idxA] <= B[idxB]):  
            merged.add(A[idxA])  
            idxA++  
        else:  
            merged.add(B[idxB])  
            idxB++  
    # grab any remaining values from the nonempty List  
    while (idxA < A.len):  
        merged.add(A[idxA])  
    while (idxB < B.len):  
        merged.add(B[idxB])  
    return merged
```

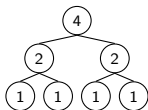
Merge Sort Pseudocode

```
def merge_sort(arr):  
    if (arr.len == 1):  
        return arr # single value sorted  
    A = arr.subList(0, arr.len/2)  
    B = arr.subList(arr.len/2, arr.len)  
    merge_sort(A)  
    merge_sort(B)  
    sorted = merge(A, B)  
    return sorted
```

Merge Sort Analysis

- Merge Sort is $O(n \lg(n))$
 - Where does the $\lg(n)$ come from?
- Recall that merge sort splits the list in half every time
 - Let's visualize the sort

Merge Sort Visual Breakdown



- The node values are the size of the list each call to the merge sort is operating on
- On the last level of the tree, there are n leaves, so there are roughly $2n$ nodes in the tree
 - This is a full tree!
 - The height is roughly $\lg(n)$
- At each level, we are doing a merge worth $O(n)$ (on level 1, we are merging $\frac{n}{2}$ twice...)
- Thus, we do approximately n operations $\lg(n)$ times:
 $O(n\lg(n))$
- $\lg(n)$ crops up frequently in divide and conquer operations

Merge Sort Summary

- Worst case (pivot is always largest or smallest): $O(n \lg(n))$
- Average case: $O(n \lg(n))$
- Not in place sort
- Not stable

- Quick Sort is like Merge Sort in that it is a divide and conquer algorithm
- Quick Sort works as follows:
 - Choose a *pivot* value
 - Partition the array so everything to the left of the pivot is less than it, and everything to the right of the pivot is greater than it
 - Recursively Quick Sort the partitions

Choosing the Pivot

- A good pivot value is the median (as it divides the array in half)
- Calculating the exact median can be slow, so approximate it
 - Choose three random numbers from the data, and choose the median of the three

Partition Pseudocode

```
def partition(arr, start, end):  
    choose pivot value  
    b = start, c = end - 1 // assumes you're sorting  
                                // to end (exclusive)  
    while (b < c):  
        while (arr[b] < pivot):  
            b++  
        while (arr[c] > pivot)  
            c--;  
        if (b <= c):  
            swap(arr[b], arr[c])  
    return index of partition (b)
```

Quick Sort Pseudocode

```
def quicksort(arr, start, end):  
    if start >= end:  
        return;  
    pivot = partition(arr, start, end);  
    quicksort(arr, start, pivot)  
    quicksort(arr, pivot+1, end)
```

Quick Sort Summary

- The partitioning scheme presented can get hairy when you have duplicate elements
- Worst case (pivot is always largest or smallest): $O(n^2)$
- Average case: $O(n \lg(n))$
- In place sort
- Not stable

Comparing Merge and Quick

Merge Sort

- Not stable
- Worst, average, and best time: $O(n \lg(n))$
- Not in place

Quick Sort

- Not stable
- Worst time: $O(n^2)$
- Average time: $O(n \lg(n))$
- Though poor worst time behavior, considered one of the most efficient sorting algorithms and is widely used
- In place sort (no new copies of data)

- All sorts we've discussed have been comparison based
 - Worst time can be no better than $O(n \lg(n))$
- What if we could sort without comparing?
- **Radix Sort** is not based on comparisons
 - Cannot be used to sort arbitrary data sets
 - Usually used for sorting unsigned integers

- Radix sort uses “bucketing”
 - Group the numbers based on individual digits
- Example: Sort the following:

3 17 12 98 101 207 8 200 45 57 73 87 41 58

Radix Sort Example

3 17 12 98 101 207 8 200 45 57 73 87 41 58

- Consider each number from left to right
- Place each number in a list based on the ones digit
- Go through the buckets in order, and rewrite the list with the numbers in order

0: 200

1: 100 41

2: 12

3: 3 73

4:

5: 45

6:

7: 17 207 57 87

8: 98 8 58

9:

200 101 41 12 3 73 45 17 207 57 87 98 8 58

Radix Sort Example

200 101 41 12 3 73 45 17 207 57 87 98 8 58

0: 200 101 3 207 8

1: 12 17

2:

3:

4: 41 45

5: 57 58

6:

7: 73

8: 87

9: 98

- Repeat the process using the tens digit
 - Single digit numbers have a 0 in the tens digit

200 101 3 207 8 12 17 41 45 57 58 73 87 98

Radix Sort Example

200 101 3 207 8 12 17 41 45 57 58 73 87 98

- Repeat the process using the hundreds digit
 - Single and double digit numbers have a 0 in the hundreds digit
- After rewriting the list, the numbers are sorted
 - No number has 4 or more digits

0: 3 8 12 17 41 45 57 58 73 87 98
1: 101
2: 200 207
3:
4:
5:
6:
7:
8:
9:

3 8 12 17 41 45 57 58 73 87 98 101 200 207

Radix Sort analysis

- Stable Sort
- Not in place sort
- Average and worst time is $O(n \log_{10}(N))$
 - N is the largest integer in the array