

CSE 373: Data Structures and Algorithms

Lecture 20: More Sorting

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Today: More sorting algorithms!

- Merge sort analysis
- Quicksort
- Bucket sort
- Radix sort

Divide and conquer

Very important technique in algorithm design

1. Divide problem into smaller parts
2. Independently solve the simpler parts
 - Think recursion
 - Or parallelism
3. Combine solution of parts to produce overall solution

Two great sorting methods are fundamentally divide-and-conquer
(Merge Sort & Quicksort)

Merge Sort

Merge Sort: repeatedly...

- Sort the left half of the elements
- Sort the right half of the elements
- Merge the two sorted halves into a sorted whole

To sort array from position l_0 to position h_i :

- If range is 1 element long, it is already sorted!
- Else:
 - Sort from l_0 to $(h_i + l_0) / 2$
 - Sort from $(h_i + l_0) / 2$ to h_i
 - Merge the two halves together

Linked lists and big data

We defined sorting over an array, but sometimes you want to sort linked lists

One approach:

- Convert to array: $O(n)$
- Sort: $O(n \log n)$
- Convert back to list: $O(n)$

$O(n^2)$ & $O(n \log n)$

Merge sort works very nicely on linked lists directly

- Heapsort and quicksort do not
- Insertion sort and selection sort do but they're slower

Merge sort is also the sort of choice for external sorting

- Linear merges minimize disk accesses
- And can leverage multiple disks to get streaming accesses

Analysis

Having defined an algorithm and argued it is correct, we should analyze its running time and space:

To sort n elements, we:

- Return immediately if $n=1$
- Else do 2 subproblems of size $n/2$ and then an $O(n)$ merge

Base case $n = \# \text{ elements}$

Recurrence relation:

$$T(1) = c_1$$

$$T(n) = c_2 n + 2T(n/2)$$

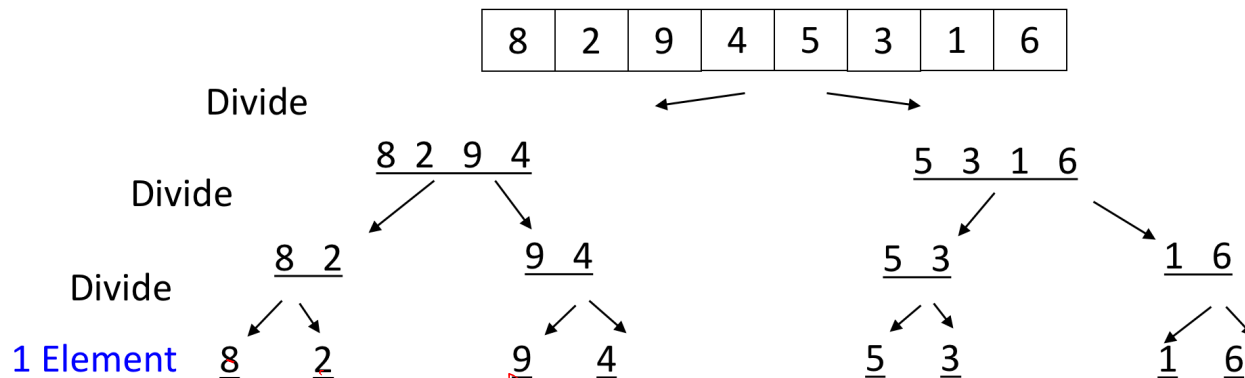
$$\leftarrow O(n \log n)$$

Analysis intuitively

This recurrence is common, you just “know” it’s $O(n \log n)$

Merge sort is relatively easy to intuit (best, worst, and average):

- The recursion “tree” will have height $\log n$
- At each level we do a *total* amount of merging equal to n



Analysis more formally

(One of the recurrence classics)

For simplicity, ignore constants (let constants be)

$$T(1) = 1$$

$$T(n) = 2T(n/2) + n$$

$$= 2(2T(n/4) + n/2) + n$$

$$= 4T(n/4) + 2n$$

$$= 4(2T(n/8) + n/4) + 2n$$

$$= 8T(n/8) + 3n$$

....

$$= 2^k T(n/2^k) + kn$$

We will continue to recurse until we reach the base case, i.e. $T(1)$ for $T(1)$, $n/2^k = 1$, i.e., $\log n = k$

So the total amount of work is $2^k T(n/2^k) + kn = 2^{\log n} T(1) + n \log n = n + n \log n = O(n \log n)$

Divide-and-Conquer Sorting

Two great sorting methods are fundamentally divide-and-conquer

1. Merge Sort:

- Sort the left half of the elements (recursively)
- Sort the right half of the elements (recursively)
- Merge the two sorted halves into a sorted whole



2. Quicksort:

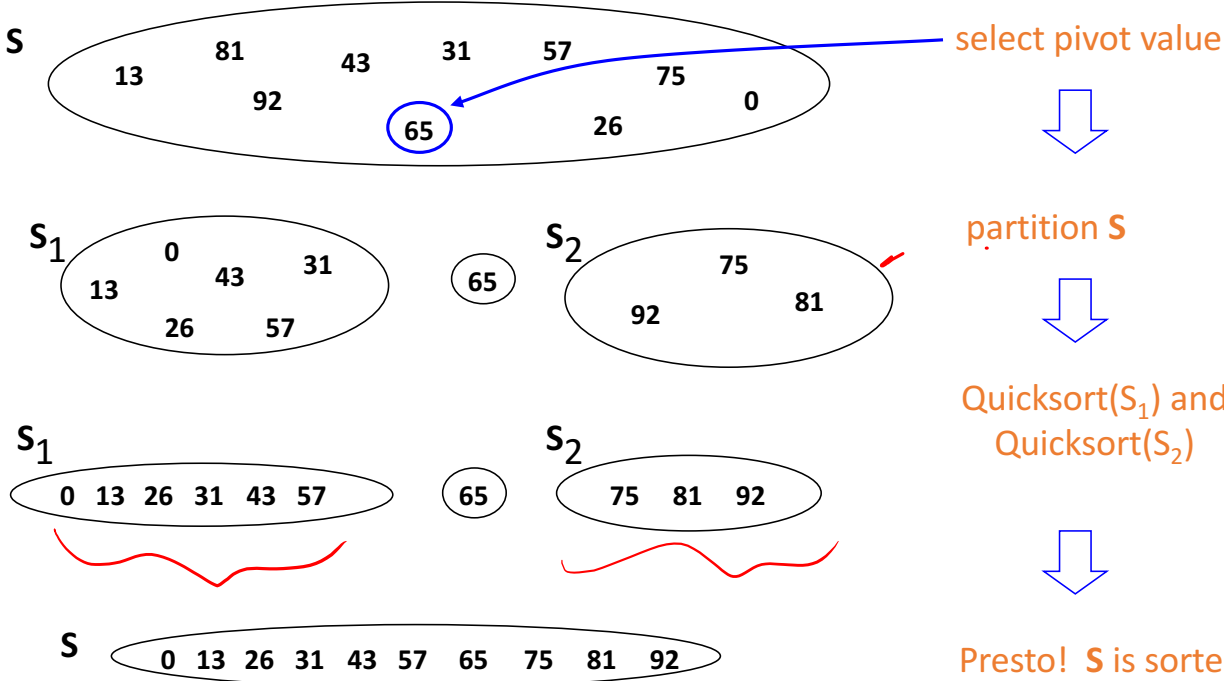
- Pick a “pivot” element
- Divide elements into “less-than pivot” and “greater-than pivot”
- Sort the two divisions (recursively on each)
- Answer is “sorted-less-than”, followed by “pivot”, followed by “sorted-greater-than”

Quicksort Overview

1. Pick a pivot element
2. Partition all the data into:
 - A. The elements less than the pivot ←
 - B. The pivot
 - C. The elements greater than the pivot ←
3. Recursively sort A and C
4. The final answer is A-B-C

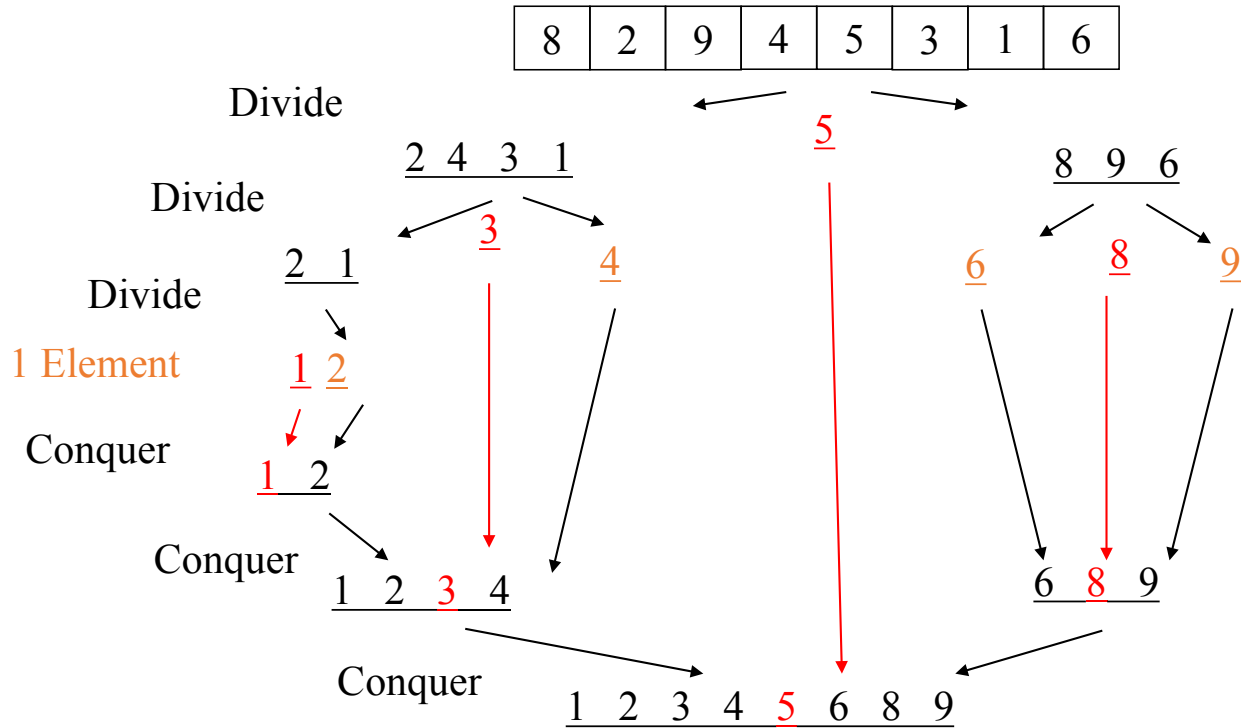
Real-world example demo time!

Think in Terms of Sets



[Weiss]

Example, Showing Recursion



Details

Have not yet explained:

- How to pick the pivot element
 - Any choice is correct: data will end up sorted
 - But as analysis will show, want the two partitions to be about *equal size*
- How to implement partitioning
 - In linear time
 - In place

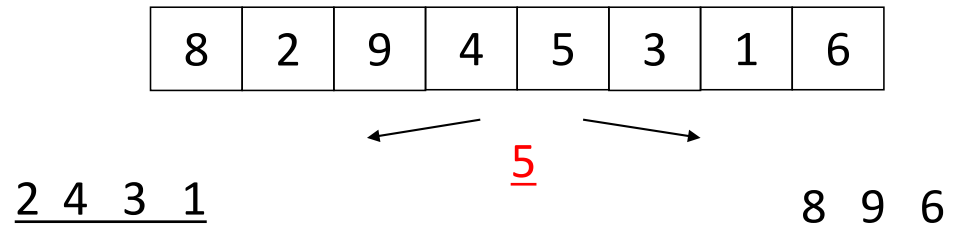
Pivots

- Best pivot?

• median

- Halve each time

• $O(n \log n)$

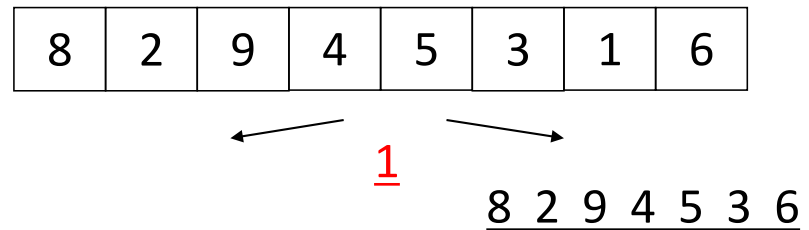


- Worst pivot?

- Greatest/least element

- Partition of size $n - 1$

• $O(n^2)$



Potential pivot rules

While sorting `arr` from `lo` to `hi-1` ...

- Pick `arr[lo]` or `arr[hi-1]`
 - Fast, but worst-case occurs with mostly sorted input
- Pick random element in the range
 - Does as well as any technique, but (pseudo)random number generation can be slow
 - Still probably the most elegant approach
- Median of 3, e.g., `arr[lo]`, `arr[hi-1]`, `arr[(hi+lo)/2]`
 - Common heuristic that tends to work well

Partitioning

Conceptually simple, but hardest part to code up correctly

- After picking pivot, need to partition in linear time in place

One approach (there are slightly fancier ones):

1. Swap pivot with `arr[lo]`
2. Use two fingers `i` and `j`, starting at `lo+1` and `hi-1`
3.

```
while (i < j)
    if (arr[j] > pivot) j--
    else if (arr[i] < pivot) i++
    else swap arr[i] with arr[j]
```
4. Swap pivot with `arr[i]` *

*skip step 4 if pivot ends up being least element


Example

- Step one: pick pivot as median of 3
 - $lo = 0, hi = 10$

0	1	2	3	4	5	6	7	8	9
8	1	4	9	0	3	5	2	7	6

- Step two: move pivot to the lo position

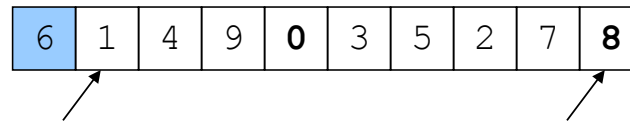
0	1	2	3	4	5	6	7	8	9
6	1	4	9	0	3	5	2	7	8



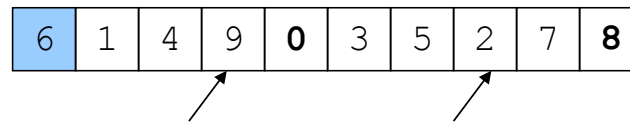
Example

Often have more than one swap during partition – this is a short example

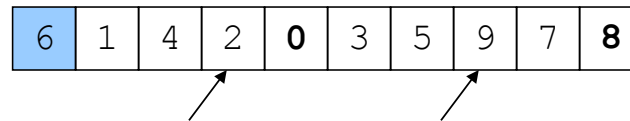
Now partition in place



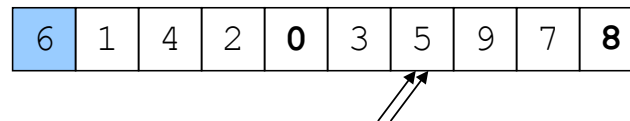
Move fingers



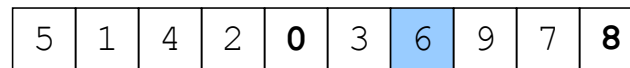
Swap



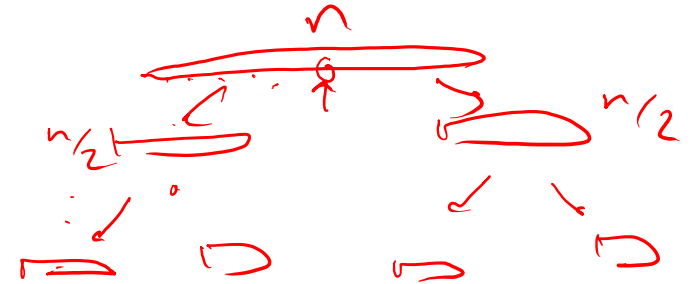
Move fingers



Move pivot



Analysis



- Best-case: Pivot is always the median

$$T(0) = T(1) = 1$$

$$T(n) = 2T(n/2) + n \text{ -- linear-time partition}$$

Same recurrence as merge sort: $O(n \log n)$

- Worst-case: Pivot is always smallest or largest element

$$T(0) = T(1) = 1$$

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

Basically same recurrence as selection sort: $O(n^2)$

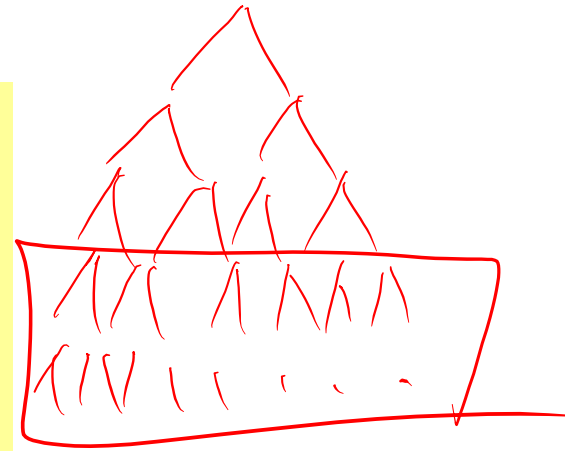
- Average-case (e.g., with random pivot)
 - $O(n \log n)$, not responsible for proof (in text)

Cutoffs

- For small n , all that recursion tends to cost more than doing a quadratic sort
 - Remember asymptotic complexity is for *really large n ($n \rightarrow \infty$)*
- Common engineering technique: switch algorithm below a **cutoff**
 - Reasonable rule of thumb: use insertion sort for $n < 10$
- Notes:
 - Could also use a cutoff for merge sort
 - Cutoffs are also the norm with parallel algorithms
 - Switch to sequential algorithm
 - None of this affects asymptotic complexity

Cutoff pseudocode

```
void quicksort(int[] arr, int lo, int hi)
{
    if(hi - lo < CUTOFF)
        insertionSort(arr, lo, hi);
    else
        ...
}
```



Notice how this cuts out the vast majority of the recursive calls

- Think of the recursive calls to quicksort as a tree
- Trims out the bottom layers of the tree

Practice with comparison sort!

A comparison sorting algorithm is operating on an array of 8 integers. After its 4th loop or recursive call, the array looks like:

4	8	11	15	42	29	18	37
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Which of these sorting algorithms can it be?

A) Heapsort

B) Merge sort

C) Insertion sort

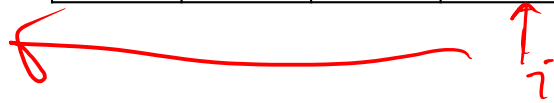
D) Quicksort using Median of 3

← no cutoff

Practice with comparison sort!

A comparison sorting algorithm is operating on an array of 8 integers. After its 4th loop or recursive call, the array looks like:

4	8	11	15	42	29	18	37
---	---	----	----	----	----	----	----



Which of these sorting algorithms can it be?

A) Heapsort

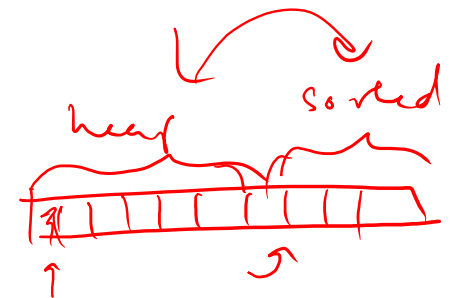
max heap (delete Max)

B) Merge sort

C) Insertion sort

D) Quicksort using Median of 3

$A[low]$, $A[high-1]$, $A[\frac{low+high}{2}]$



How Fast Can We Sort?

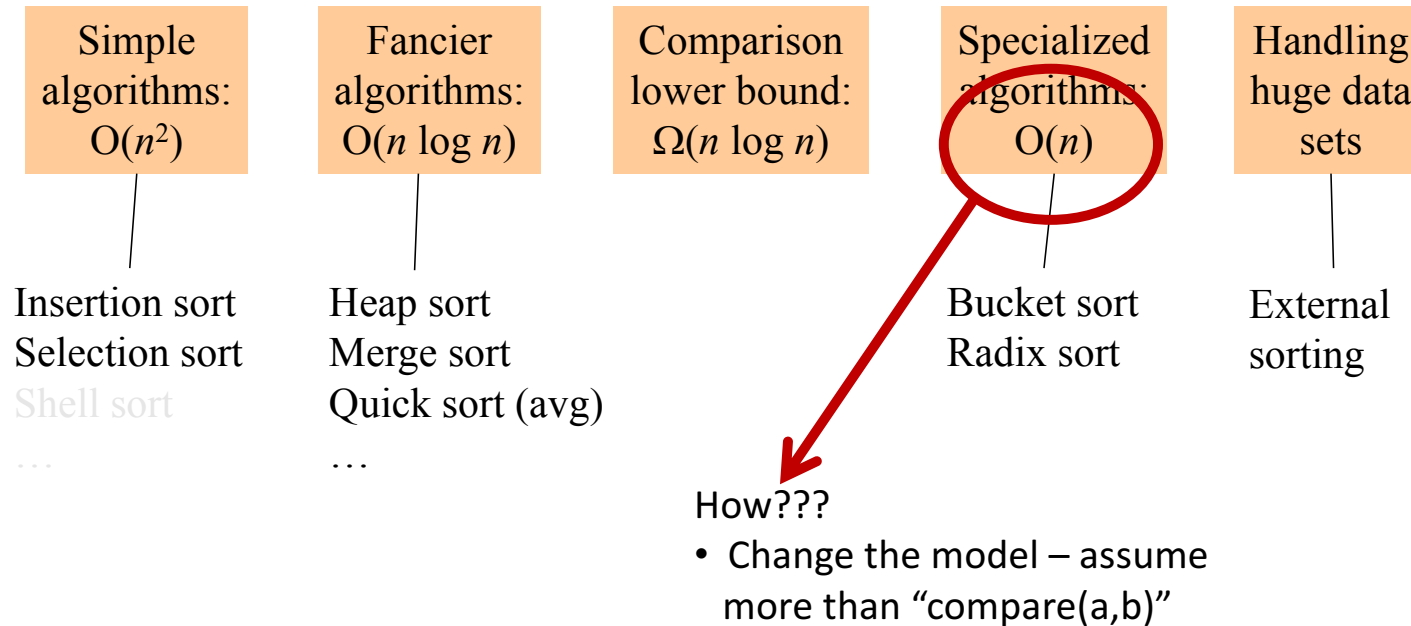
- Heapsort & mergesort have $O(n \log n)$ worst-case running time
- Quicksort has $O(n \log n)$ average-case running time
- These bounds are all tight, actually $\Theta(n \log n)$
- Comparison sorting in general is $\Omega(n \log n)$
 - An amazing computer-science result: proves all the clever programming in the world cannot comparison-sort in linear time

lower bound!

~~$O(n)$~~


The Big Picture

Surprising amount of juicy computer science: 2-3 lectures...



Bucket Sort (a.k.a. BinSort)

- If all values to be sorted are *known* to be integers between 1 and *K* (or any small range):
 - Create an array of size *K*
 - Put each element in its proper **bucket (a.k.a. bin)**
 - *If* data is only integers, no need to store more than a *count* of how times that bucket has been used
- Output result via linear pass through array of buckets



count array	
1	
2	
3	
4	
5	

- Example:

K=5

input (~~5~~, ~~1~~, ~~3~~, ~~4~~, ~~3~~, ~~2~~, ~~1~~, ~~1~~, ~~5~~, ~~4~~, ~~5~~)

output 1, 1, 1, 2, 3, 3, 4, 4, 5, 5, 5

Analyzing Bucket Sort

$n = \# \text{ elements}$
 $K = \# \text{ buckets}$

- Overall: $O(n+K)$
 - Linear in n , but also linear in K
 - $\Omega(n \log n)$ lower bound does not apply because this is not a comparison sort
- Good when K is smaller (or not much larger) than n
 - We don't spend time doing comparisons of duplicates
- Bad when K is much larger than n
 - Wasted space; wasted time during linear $O(K)$ pass
- For data in addition to integer keys, use list at each bucket

Bucket Sort with Data

- Most real lists aren't just keys; we have data
- Each bucket is a list (say, linked list)
- To add to a bucket, insert in $O(1)$ (at beginning, or keep pointer to last element)

Example: spice level; scale 1-5;

1 = mild, 5 = very spicy

Input=

5: Habanero

3: Jalapeño

5: Ghost pepper

1: Bell pepper

count array	
1	
2	
3	
4	
5	

→ Bell

→ Jalapeño

→ Habanero → Ghost

- Result: Bell, Jalapeño, Habanero, Ghost
- Easy to keep 'stable'; Habanero still before Ghost pepper

Interactive Visualizations

Comparison Sort (including quicksort):

- <http://www.cs.usfca.edu/~galles/visualization/ComparisonSort.html>

Bucket Sort:

- <http://www.cs.usfca.edu/~galles/visualization/BucketSort.html>
- <http://www.cs.usfca.edu/~galles/visualization/CountingSort.html>

Radix Sort:

- <http://www.cs.usfca.edu/~galles/visualization/RadixSort.html>