

# CMPT 280

## Topic 23: Efficient Sorting Algorithms

Mark G. Eramian

University of Saskatchewan

# References

- Textbook, Chapter 23

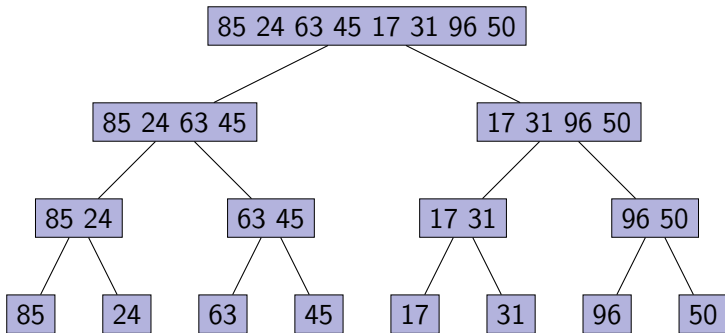
# Divide And Conquer Sorts

- Merge Sort and Quick Sort solve the sorting problem using the *divide-and-conquer* approach.
- Basic divide-and-conquer idea:
  - **Divide:** Divide the problem into smaller sub-problems.
  - **Recurse:** Recursively try to solve the sub-problems. When we reach a level where the sub-problem is small enough, solve it quickly.
  - **Conquer:** Take the solutions of the subproblems and combine them into a solution of the original problem.

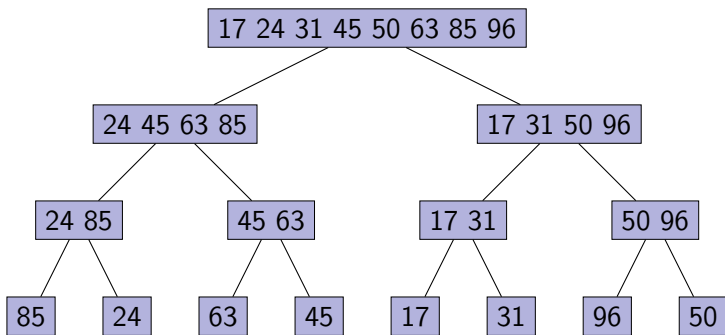
# Merge Sort

- We assume that we will be sorting a sequence of elements  $S$  with  $n$  elements.
- The merge sort uses the following divide and conquer approach:
  - **Divide:** If  $S$  has zero or one elements, return  $S$  immediately, it is already sorted. Otherwise, divide  $S$  into  $S_1$  and  $S_2$  such that  $S_1$  contains the first  $\frac{n}{2}$  elements of  $S$  and  $S_2$  contains the remaining  $\frac{n}{2}$  (rounding up and down respectively).
  - **Recurse:** Recursively sort the sequences  $S_1$  and  $S_2$ .
  - **Conquer:** Put elements back into  $S$  by merging the sorted sequences  $S_1$  and  $S_2$ .
- We can visualize what is happening with a binary tree  $T$  called a *merge-sort-tree*.

## A Mergesort Tree (Divide Step) – Input Sequences



## A Mergesort Tree (Conquer Step) – Output Sequences



Note that the height of the merge sort tree is  $\lceil \log_2 n \rceil$ .

## Exercise 1

This is the array of the previous example, just before the conquer step of the root of the merge sort tree. Each half contains a sorted sequence.

- a) How is merge called to merge the two sequences?
- b) Trace through merge step-by-step to show how the merging happens.

0	1	2	3	4	5	6	7
24	45	63	85	17	31	50	96

# Exercise 1

```
1  /** Merge start1st through start2nd-1 with start2nd through end2nd in
2      array inVec using temp as a temporary array */
3  private void merge(int[] inVec, int[] temp, int start1st,
4                    int start2nd, int end2nd)
5  {
6      int cur1 = start1st; // index of the current item in first half
7      int cur2 = start2nd; // index of the current item in first half
8
9      int cur3 = 0; // index of the next location of temp
10     // While neither subarray is empty...
11     while ((cur1 < start2nd) && (cur2 <= end2nd))
12         // Copy the smaller item from the two sequences to temp
13         if (inVec[cur1] <= inVec[cur2])
14             // Note: increments happen after assignment
15             temp[cur3++] = inVec[cur1++];
16         else
17             temp[cur3++] = inVec[cur2++];
18
19     while (cur1 < start2nd) // copy remainder (if any) of first subarray to temp
20         temp[cur3++] = inVec[cur1++];
21
22     while (cur2 <= end2nd) // copy remainder (if any) of second subarray to temp
23         temp[cur3++] = inVec[cur2++];
24
25     // Copy the temp array back to inVec[start1st...end2nd]
26     cur1 = start1st;
27     cur3 = 0;
28     while (cur1 <= end2nd) // copy items from temp back into inVec
29         inVec[cur1++] = temp[cur3++];
30 }
```



## Analysis of Merge Operation

- Let  $n_1$  and  $n_2$  be the number of elements in  $S_1$  and  $S_2$ .
- Assume insertions and deletions from first and last positions of  $S_1$  and  $S_2$  are  $O(1)$ .
- In each iteration of all three while loops, an element is removed from either  $S_1$  or  $S_2$ .
- No elements are ever inserted into  $S_1$  or  $S_2$ .
- Number of loop iterations in all three loops must therefore be  $n_1 + n_2$ .
- Running time is thus  $O(n_1 + n_2)$  – linear time!.

## Exercise 2

For the following array of elements, draw the merge sort tree showing the input sequences at each recursive call (after the divide step) and the output sequences at each recursive call (after the conquer step)

0	1	2	3	4	5	6	7	8	9	10	11
83	42	5	33	91	62	29	0	73	48	11	12

# Analysis of Merge Sort

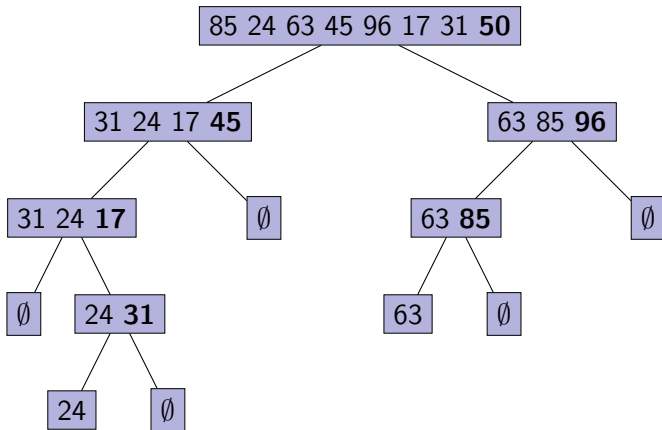
- Analysis refers to a merge-sort-tree  $T$ .
- Define *time spent at node  $v$*  to be the time taken for the divide step, the conquer step, but not the recurse step.
- Both divide and conquer steps run in at most linear time. If  $i$  is the depth of  $v$ , then the size of the sequence at  $v$  contains at most  $\frac{n}{2^i}$  elements, thus time spent at node  $v$  is  $O(\frac{n}{2^i})$ .
- There are at most  $2^i$  nodes at depth  $i$ , so time spent at all nodes of depth  $i$  is  $O(2^i \cdot \frac{n}{2^i}) = O(n)$ .
- Since the height of  $T$  is  $\lceil \log n \rceil$ , time spent at all  $\lceil \log n \rceil$  levels is  $O(n \log n)$ .

# Quick Sort

- Hard work done **before** recursive calls.
- Idea of quick sort of a sequence  $S$ :
  - **Divide:** If  $S$  has at least two elements, select some element  $x$  from  $S$ , called the *pivot*. Remove each element from  $S$  and place them in one of three sequences:
    - $L$  (for elements less than  $x$ )
    - $E$  (for elements equal to  $x$ )
    - $G$  (for elements greater than  $x$ )
  - **Recurse:** Recursively sort  $L$  and  $G$ .
  - **Conquer:** Put  $S$  back together by concatenating the (already sorted) elements of  $L$ ,  $E$ , and  $G$ .

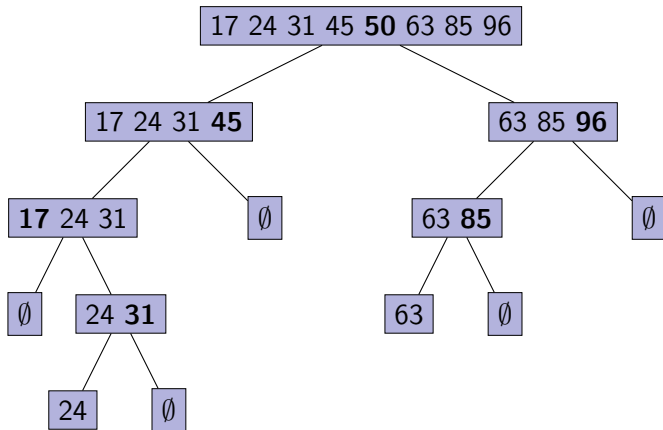
## A Quick Sort Tree (Divide Step) – Input Sequences

This quick sort tree shows the input sequences processed by each recursive call. The last element of a sequence is used as the pivot; these are shown in bold. Note: order of child input sequences are affected by parent's partitioning!



## A Quick Sort Tree (Conquer Step) – Output Sequences

This quick sort tree shows the output sequences of each recursive call (after the conquer step). Again, pivots are in bold.



## Exercise 3

The work of quick sort is the partition algorithm that performs the divide step. The following array is the input to the first recursive call to quick sort at the root of the quick sort tree.

- a) How is the partition algorithm called to partition the array?
- b) Trace through the partition algorithm to show how the array is partitioned in-place.

0	1	2	3	4	5	6	7
19	42	77	71	41	9	83	68

## Exercise 3

```
1 private int partition(int[ ] inVec, int start, int finish) {
2     int partitionItem = inVec[finish]; // item that will be the pivot
3     int l = start-1; // initial lower offset
4     int r = finish; // initial upper offset
5     while (l < r)
6     {
7         l++;
8         // move l right, passing "small" items
9         while ((l < r) && (inVec[l] <= partitionItem))
10             l++;
11         r--;
12         // move r left, passing "large" items
13         while ((l <= r) && (inVec[r] >= partitionItem))
14             r--;
15
16         // At this point, r is the offset of a "small" item
17         // and l is the offset of a "large" item.
18         if (l < r) // If l and r have *not* yet crossed...
19         {
20             // swap the large item at offset l with the
21             // small item at offset r
22             int temp = inVec[l];
23             inVec[l] = inVec[r];
24             inVec[r] = temp;
25         }
26     }
27     // put item l in the last offset of the subarray, and
28     // partitionItem (the pivot) in position l
29     inVec[finish] = inVec[l];
30     inVec[l] = partitionItem;
31     return l;
32 }
```



## Timing Analysis of partition

- Worst case: pivot is the median, and the rest of the array is in reverse order.
- Each time through the outer while loop,  $l$  moves exactly one position to the right, and  $r$  moves exactly one position to the left.
- Swaps are  $O(1)$  so each iteration of the while loop is  $O(1)$  in the worst case.
- The outer while loop will execute  $n/2$  times before  $l \geq r$ .
- Total cost of while loop:  $O(n)$
- Initialization and final swap are  $O(1)$ , so total algorithm cost is  $O(n) + O(1)$  or just  $O(n)$ .

## Exercise 4

For the following array of elements, draw the quick sort tree showing the input sequences at each recursive call (after the divide step) and the output sequences at each recursive call (after the conquer step)

0	1	2	3	4	5	6	7	8	9	10	11
83	42	5	33	91	62	29	0	73	48	11	12

# Analysis of Quick-Sort

- Let  $s(v)$  be the size of the sequence to be sorted at node  $v$ .
- The partitioning step is linear time:  $O(s(v))$ .
- The conquer step (concatenating sequences) is also at most linear:  $O(s(v))$
- Thus, the time spent at each node in the quick-sort-tree is at most  $O(2s(v))$ , or, more simply  $O(s(v))$ .
- Worst case is when the sequence is **already sorted**!
- What happens to our quick-sort-tree visualization if the array is already sorted?

## Analysis of Quick Sort

- Let  $s_i$  denote the sum of the input sizes of nodes at depth  $i$ . Clearly  $s_0 = n$ .
- In the worst case,  $s_i = n - i$ .
- Thus the worst case running time is

$$O\left(\sum_{i=0}^{n-1} s_i\right) \text{ which is } O\left(\sum_{i=0}^{n-1} (n - i)\right) \text{ which is } O\left(\sum_{i=1}^n i\right)$$

- We know that  $\sum_{i=1}^n i$  is  $O(n^2)$ , thus quick sort runs in  $O(n^2)$  in the worst case.
- However, in the best case quick sort runs in  $O(n \log n)$  time.
- What is the best case??

# Sorting Algorithms

## Quick Sort - Remarks

- Quick Sort is not stable.
- In practice, quick sort is often a little faster than merge sort.
- Merge sort, however, gives you a guaranteed worst case time complexity of  $O(n \log n)$ . Quick sort does not. The worst case is still  $O(n^2)$ .
- Can we modify quick sort to get a worst case bound of  $O(n \log n)$ ?

# Sorting Algorithms

Quick Sort -  $O(n \log n)$  guarantee?

- If we always used as a pivot the median of the sequence being partitioned, we would get a balanced quick sort tree and a guarantee of  $O(n \log n)$ .
- But... the straightforward way to find the median is to sort the items and select the middle one – **this does not help!**.
- It is possible to find the median of a sequence in as quickly as  $O(n)$  time. But this algorithm itself is a modification of quick sort. It would give the asymptotic  $O(n \log n)$  guarantee, but it would probably increase the constant so that merge sort was always faster in practice.

# Sorting Algorithms

Quick Sort -  $O(n \log n)$  guarantee?

- Approximate the median: use the median of the first 3 items as the pivot. Reduces the likelihood of a bad pivot, but still not a total guarantee.
- Choose the pivot at random: Usually does well, still no perfect guarantee. Also the cost of generating lots of random numbers can be higher than the overall savings obtained by reducing the likelihood of bad partitioning.

# Heap Sort

The basics of Heap Sort:

- Heapify the input array.
- Sequentially delete items from the heap, obtaining items in reverse order, moving each one into position as it is obtained. Re-heapify after each deletion as normal.



## Exercise 5

Trace through the heapify algorithm showing, step-by-step, how the following array to be sorted is converted into an arrayed heap.

0	1	2	3	4	5	6	7	8	9	10	11
83	42	5	33	91	62	29	0	73	48	11	12

## Exercise 5

```
1  Algoirthm heapify(data)
2  data - array of elements to convert to a heap
3
4  n = data.length
5  // We can start at i = n/2-1 because larger offsets are guaranteed
6  // to be leaves by the properties of arrayed heaps.
7  for i = n/2-1 to 0
8      // Make data[i] the root of a valid heap
9      // by swapping data[i] with children repeatedly as needed.
10     moveDown(data, i, n-1)
11
12 Algoirthm moveDown( data, first, last )
13 data - array of elements to be converted to heap
14 first - offset of element to process
15 last - offset of last element in the 'data' array
16
17 // select the left child of offset 'first'
18 largest = 2 * first + 1;
19 while( largest <= last )
20     // if a right child exists and its item is bigger than data[largest]...
21     if( largest < last and data [largest] < data[largest+1] )
22         largest++ // select the right child instead
23
24     // If item at offset 'first' is smaller than its larger child
25     // it does not have the heap property, so swap it with its larger child.
26     if( data[first] < data[largest] )
27         swap( data[first], data[largest] ) // swap it
28         first = largest // follow the item down
29         largest = 2 * first + 1 // select its new left child
30     else
31         // cause loop to end -- data[first] is the root of a valid heap
32         largest = last + 1
```

## Exercise 6

Beginning with the solution to Exercise 5, show step-by-step how the heap sort is completed by tracing through the heap sort algorithm.

```
1  Algorithm heapSort(data)
2  data - Array of items to be sorted
3
4  heapify(data)
5  i = data.length-1
6  while( i > 0 )
7      // Move the next largest item into sorted position
8      swap(data[0], data[i])
9      i = i - 1;
10     // Re-heapify by moving the new root down.
11     moveDown(data, 0, i)
```

# Efficiency of Sorts

	Worst Case	Best Case
Selection Sort	$O(n^2)$	$O(n^2)$
Insertion Sort	$O(n^2)$ (array in reverse order)	$O(n)$ (array already sorted)
Quick Sort*	$O(n^2)$ (array already sorted)	$O(n \log n)$ (by luck, pivot is always the median)
Merge Sort	$O(n \log n)$	$O(n \log n)$
Tree Sort**	$O(n \log n)$	$O(n \log n)$
Heap sort	$O(n \log n)$	$O(n \log n)$

\* Quick sort is  $O(n \log n)$  in the average case.

## Next Class

- **Next class reading:** Chapter 24: Linear-time Sorting Algorithms