

# Introduction to Algorithms

**George Voutsadakis<sup>1</sup>**

<sup>1</sup>Mathematics and Computer Science  
Lake Superior State University

LSSU Math 400

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  - Description of Quicksort
  - Performance of Quicksort
  - A Randomized Version of Quicksort

## Subsection 1

### Description of Quicksort

# Description of QUICKSORT

- Quicksort, like merge sort, applies the divide-and-conquer paradigm.
- The three-step divide-and-conquer process for sorting a typical subarray  $A[p \dots r]$ :
  - **Divide**: Partition (rearrange) the array  $A[p \dots r]$  into two (possibly empty) subarrays  $A[p \dots q - 1]$  and  $A[q + 1 \dots r]$ , such that each element of  $A[p \dots q - 1]$  is less than or equal to  $A[q]$ , which is, in turn, less than or equal to each element of  $A[q + 1 \dots r]$ . Compute the index  $q$  as part of this partitioning procedure.
  - **Conquer**: Sort the two subarrays  $A[p \dots q - 1]$  and  $A[q + 1 \dots r]$  by recursive calls to quicksort.
  - **Combine**: Because the subarrays are already sorted, no work is needed to combine them: The entire array  $A[p \dots r]$  is now sorted.

# The Procedure QUICKSORT

## QUICKSORT( $A, p, r$ )

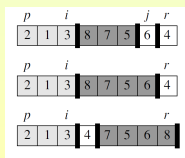
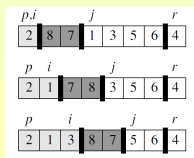
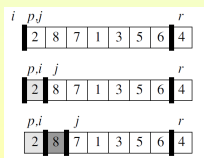
1. **if**  $p < r$
2.      $q = \text{PARTITION}(A, p, r)$
3.     QUICKSORT( $A, p, q - 1$ )
4.     QUICKSORT( $A, q + 1, r$ )

- To sort an array  $A$ , the initial call is QUICKSORT( $A, 1, A.\text{length}$ ).
- The key to the algorithm is the PARTITION procedure, which rearranges the subarray  $A[p \dots r]$  in place.

# The PARTITION Procedure

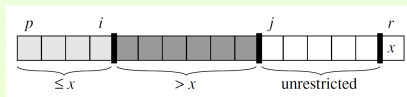
PARTITION( $A, p, r$ )

1.  $x = A[r]$
2.  $i = p - 1$
3. for  $j = p$  to  $r - 1$
4.   if  $A[j] \leq x$
5.      $i = i + 1$
6.     exchange  $A[i]$  with  $A[j]$
7. exchange  $A[i + 1]$  with  $A[r]$
8. return  $i + 1$



# How PARTITION Works

- PARTITION always selects an element  $x = A[r]$  as a **pivot** element around which to partition the subarray  $A[p \dots r]$ .
- As the procedure runs, it partitions the array into four (possibly empty) regions that satisfy a loop invariant:



For any array index  $k$ ,

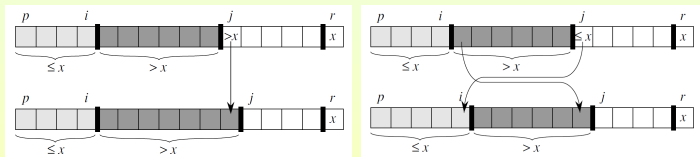
1. If  $p \leq k \leq i$ , then  $A[k] \leq x$ .
2. If  $i + 1 \leq k \leq j - 1$ , then  $A[k] > x$ .
3. If  $k = r$ , then  $A[k] = x$ .

The indices between  $j$  and  $r - 1$  are not covered by any of the three cases, and these values have no particular relationship to the pivot  $x$ .

- We need to show that this loop invariant (a) is true prior to the first iteration, (b) is maintained by each iteration, (c) provides a helps in showing correctness when the loop terminates.

# Initialization and Maintenance

- Initialization:** Prior to the first iteration of the loop,  $i = p - 1$ , and  $j = p$ . There are no values between  $p$  and  $i$ , and no values between  $i + 1$  and  $j - 1$ , so the first two conditions of the loop invariant are trivially satisfied. The assignment in Line 1 satisfies the third condition.
- Maintenance:** There are two cases to consider, depending on the outcome of the test in Line 4:

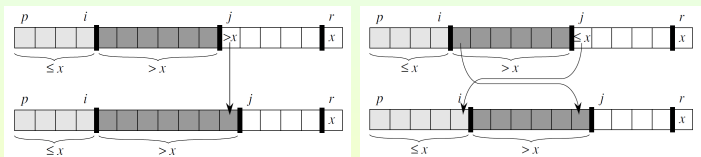


- If  $A[j] > x$ , the only action in the loop is to increment  $j$ . After  $j$  is incremented, Condition 2 holds for  $A[j - 1]$  and all other entries remain unchanged.



# Maintenance (Cont'd)

- Maintenance:** There are two cases to consider, depending on the outcome of the test in Line 4:



- If  $A[j] \leq x$ ,  $i$  is incremented,  $A[i]$  and  $A[j]$  are swapped, and then  $j$  is incremented. Because of the swap, we now have that  $A[i] \leq x$ , and Condition 1 is satisfied. Similarly, we also have that  $A[j - 1] > x$ , since the item that was swapped into  $A[j - 1]$  is, by the loop invariant, greater than  $x$ .

# Termination and Running Time

- **Termination:** At termination,  $j = r$ . Therefore, every entry in the array is in one of the three sets described by the invariant, and we have partitioned the values in the array into three sets:
  - those less than or equal to  $x$ ;
  - those greater than  $x$ ;
  - a singleton set containing  $x$ .
- The final two lines of `PARTITION` move the pivot element into its place in the middle of the array by swapping it with the leftmost element that is greater than  $x$ .

The output of `PARTITION` now satisfies the specifications given for the divide step.

- The running time of `PARTITION` on the subarray  $A[p \dots r]$  is  $\Theta(n)$ , where  $n = r - p + 1$ .

## Subsection 2

### Performance of Quicksort

# Worst-Case Partitioning

- The worst-case behavior for quicksort occurs when the partitioning produces one subproblem with  $n - 1$  and one with 0 elements. Let us assume that this unbalanced partitioning arises in each recursive call.
  - The partitioning costs  $\Theta(n)$  time.
  - A recursive call on an array of size 0 takes  $T(0) = \Theta(1)$ .

Thus, the recurrence for the running time is

$$T(n) = T(n - 1) + T(0) + \Theta(n) = T(n - 1) + \Theta(n).$$

- Using the substitution method we can prove that the recurrence  $T(n) = T(n - 1) + \Theta(n)$  has the solution

$$T(n) = \Theta(n^2).$$

- Thus, if the partitioning is maximally unbalanced at every recursive level of the algorithm, the running time is  $\Theta(n^2)$ .

# Best-Case Partitioning

- In the most even possible split, PARTITION produces two subproblems, each of size no more than  $\frac{n}{2}$ , since one is of size  $\lfloor \frac{n}{2} \rfloor$  and one of size  $\lceil \frac{n}{2} \rceil - 1$ .

In this case, quicksort runs much faster.

The recurrence for the running time is then

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

where we tolerate the sloppiness from ignoring the floor and ceiling and from subtracting 1.

- This recurrence has the solution

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

- By equally balancing the two sides of the partition at every level of the recursion, we get an asymptotically faster algorithm.

# Balanced Partitioning

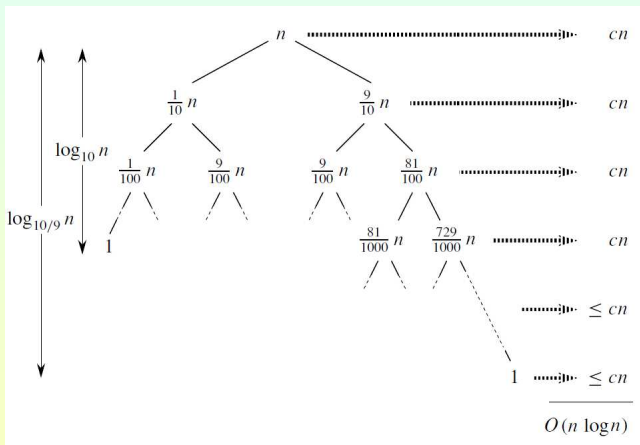
- The average-case running time of quicksort is much closer to the best case than to the worst case.
- The key to understand why is to understand **how the balance of the partitioning is reflected in the recurrence** for the running time.
- Suppose, for example, that the partitioning algorithm always produces a 9-to-1 proportional split, which seems quite unbalanced. We then obtain the recurrence

$$T(n) = T\left(\frac{9n}{10}\right) + T\left(\frac{n}{10}\right) + cn$$

on the running time of quicksort, where we have explicitly included the constant  $c$  hidden in the  $\Theta(n)$  term.

Every level of the tree has cost  $cn$ , until a boundary condition is reached at depth  $\log_{10} n = \Theta(\log n)$ . Then the levels have cost at most  $cn$ . The recursion terminates at depth  $\log_{10/9} n = \Theta(\log n)$ . The total cost of quicksort is therefore  $O(n \log n)$ .

# Balanced Partitioning (Cont'd)



- Even a 99-to-1 split yields an  $O(n \log n)$  running time, since any split of constant proportionality yields a recursion tree of depth  $\Theta(\log n)$ , where the cost at each level is  $O(n)$ .

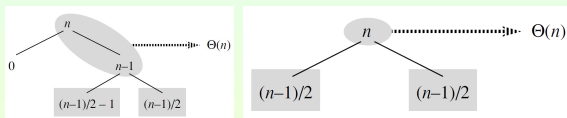
# Intuition for the Average Case

- To develop a clear notion of the average case for quicksort, we must make an assumption about **how frequently we expect to encounter the various inputs**.
- The behavior of quicksort is determined by the relative ordering of the values in the array and not by the particular values.
- We will assume for now that **all permutations of the input numbers are equally likely**.
- When we run quicksort on a random input array, it is **unlikely that the partitioning always happens in the same way at every level**: We expect that some of the splits will be reasonably well balanced (“good”) and that some will be fairly unbalanced (“bad”).
- In a recursion tree for an average-case execution of PARTITION, the good and bad splits are distributed randomly throughout the tree.



# Intuition for the Average Case (Cont'd)

- Suppose that the good and bad splits alternate levels, the good splits are best-case splits and the bad splits are worst-case splits:



- At the root of the tree, the cost is  $n$  for partitioning, and the subarrays produced have sizes  $n - 1$  and  $0$ : the worst case.
- At the next level, the subarray of size  $n - 1$  is best-case partitioned into subarrays of size  $\frac{n-1}{2} - 1$  and  $\frac{n-1}{2}$ .
- If the boundary-condition cost is  $1$  for the subarray of size  $0$ , the combination produces three subarrays of sizes  $0$ ,  $\frac{n-1}{2} - 1$  and  $\frac{n-1}{2}$  at a combined partitioning cost of  $\Theta(n) + \Theta(n - 1) = \Theta(n)$ .
- This situation is no worse than a single level of balanced partitioning that produces two subarrays of size  $\frac{n-1}{2}$ , at a cost of  $\Theta(n)$ !

## Subsection 3

# A Randomized Version of Quicksort

# Introducing Random Sampling in Quicksort

- We employ a randomization technique, called **random sampling**, to analyze a randomized version of quicksort.
- Instead of always using  $A[r]$  as the pivot, we will select a randomly chosen element from the subarray  $A[p \dots r]$ .
- We do so by first exchanging element  $A[r]$  with an element chosen at random from  $A[p \dots r]$ .
- By randomly sampling the range  $p, \dots, r$ , we ensure that the pivot element  $x = A[r]$  is equally likely to be any of the  $r - p + 1$  elements in the subarray.
- Because we randomly choose the pivot element, we expect the split of the input array to be reasonably well balanced on average.

# RANDOMIZEDQUICKSORT

- The changes to PARTITION and QUICKSORT are small.
  - In the new partition procedure, we simply implement the swap before actually partitioning:

## RANDOMIZEDPARTITION( $A, p, r$ )

1.  $i = \text{RANDOM}(p, r)$
2. exchange  $A[r]$  with  $A[i]$
3. return PARTITION( $A, p, r$ )

- The new quicksort calls RANDOMIZEDPARTITION in place of PARTITION:

## RANDOMIZEDQUICKSORT( $A, p, r$ )

1. if  $p < r$
2.  $q = \text{RANDOMIZEDPARTITION}(A, p, r)$
3. RANDOMIZEDQUICKSORT( $A, p, q - 1$ )
4. RANDOMIZEDQUICKSORT( $A, q + 1, r$ )