

# Quicksort

CS624 — Analysis of Algorithms

September 25, 2024

# Sorting, Revisited

We have seen several **sorting** algorithms so far:

- ▶ Insertion Sort (incremental)
- ▶ Merge Sort (**divide and conquer**, all work in *combine* step)
- ▶ Heap Sort

Is there a **divide and conquer** algorithm for **sorting** that does all of the work in the *divide* step instead?

Is there a **divide and conquer** algorithm for **sorting** that does all of the work in the *divide* step instead?

- ▶ Let's assume there are two **sorting** sub-problems.
- ▶ If all the work is in *divide*, then *combine* must be trivial, such as just concatenating **sorted** sub-arrays.
- ▶ For concatenation to work, one sub-array must be ordered entirely before the other sub-array.
- ▶ So our *divide* step must be to **partition** the original array such that every element of the first part is  $\leq$  every element of the second part.

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**Algorithm 1** Quicksort( $A, p, r$ )

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**Ensure:**  $A[p .. r]$  is sorted

- 1: **if**  $p < r$  **then**
  - 2:    $q \leftarrow \text{Partition}(A, p, r)$
  - 3:   Quicksort( $A, p, q - 1$ )
  - 4:   Quicksort( $A, q + 1, r$ )
  - 5: **end if**
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The Partition procedure picks an element called the “pivot” and breaks the array into three parts:  $\leq$ ,  $=$ ,  $>$  the pivot.

After `Partition` has been called the following are true:

1.  $p \leq q \leq r$ .
2. The number  $A[q]$ , the pivot, is in its final position. It will never be moved again.
3. If  $i < q$ , then  $A[i] \leq A[q]$ ,  
and if  $i > q$ , then  $A[i] > A[q]$ .

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**Algorithm 2** Partition( $A, p, r$ )

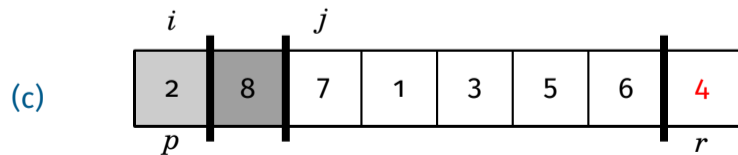
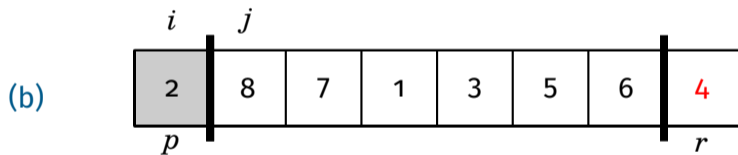
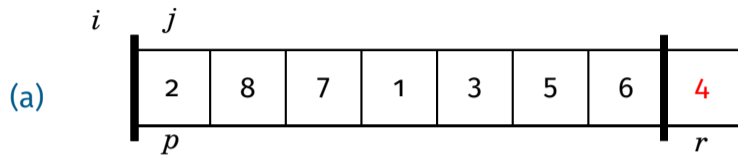
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**Ensure:** Let  $q = \mathbf{result}$ .  $A[p .. q - 1] \leq A[q] < A[q + 1 .. r]$ ,  $p \leq q \leq r$

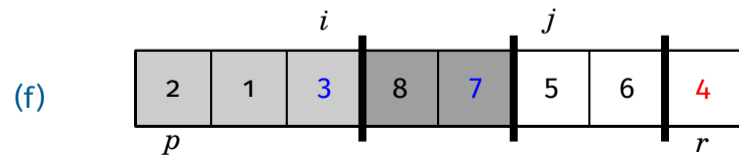
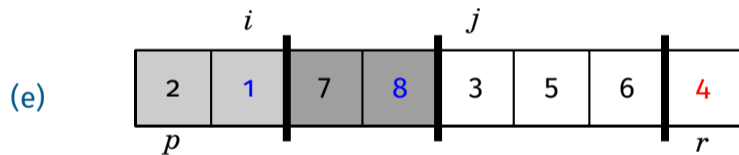
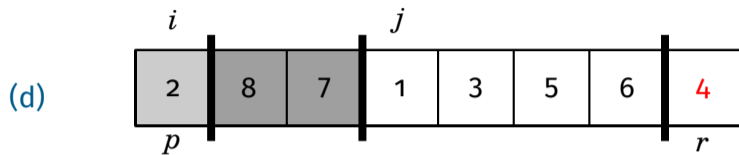
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1:  $x \leftarrow A[r]$            //  $x$  is the "pivot"
2:  $i \leftarrow p - 1$        //  $i$  maintains the "left-right boundary"
3: for  $j \leftarrow p$  to  $r - 1$  do
4:   if  $A[j] \leq x$  then
5:      $i \leftarrow i + 1$ 
6:     exchange  $A[i] \leftrightarrow A[j]$ 
7:   end if
8: end for
9: exchange  $A[i + 1] \leftrightarrow A[r]$ 
10: return  $i + 1$ 
```

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# Example: Partition

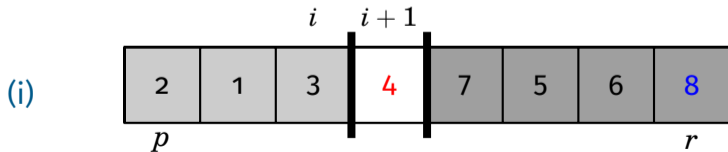
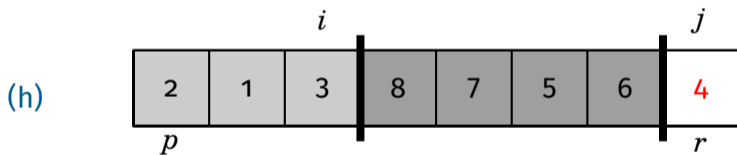
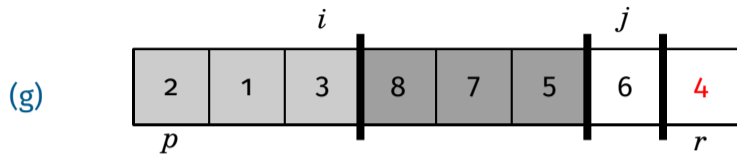


# Example: Partition

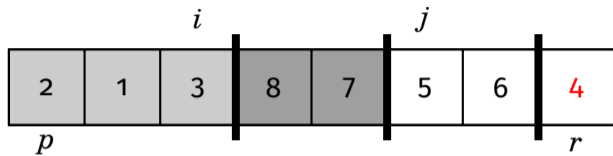




# Example: Partition



# Partition, Proof of Correctness



## Loop Invariant (Partition)

At the beginning of each iteration:

- ▶  $A[p .. i]$  are known to be  $\leq pivot$ .
- ▶  $A[i + 1 .. j - 1]$  are known to be  $> pivot$ .
- ▶  $A[j, r - 1]$  not yet examined.
- ▶  $A[r]$  is the pivot.
- ▶  $p - 1 \leq i < j$

## Lemma (Partition correctness)

Let  $q = \text{Partition}(A, p, r)$ . Then afterwards,

- ▶  $p \leq q \leq r$
- ▶  $A[p .. q - 1] \leq A[q] < A[q + 1 .. r]$

LI :  $A[p .. i] \leq \text{pivot}, A[i + 1 .. j - 1] > \text{pivot}, p - 1 \leq i < j$

## Proof.

### Initialization:

- ▶ At the beginning,  $i = p - 1$  and  $j = p$ . Both array ranges simplify to  $A[p .. p - 1]$  and  $A[p .. p - 1]$ , empty, so LI trivially holds.

# Partition, Proof of Correctness

LI :  $A[p .. i] \leq pivot$ ,  $A[i + 1 .. j - 1] > pivot$ ,  $p - 1 \leq i < j$

## Proof.

### Maintenance:

- ▶ Assume LI is true at the start of some  $j$  loop.  
In particular:  $A[p .. i] \leq pivot$  and  $A[i + 1 .. j - 1] > pivot$ .
- ▶ We must show that the execution of the loop body makes LI true for the next  $j$  value,  $j + 1$ . There are two cases:
  1. Case  $A[j] \leq pivot$ : (next page)
  2. Case  $A[j] > pivot$ : We don't move it. The  $\leq$  range stays the same, and  $A[j]$  gets absorbed into the  $>$  range, and now  $A[i + 1 .. (j + 1) - 1] > pivot$ , so the LI holds for  $j + 1$ .

# Partition, Proof of Correctness

LI :  $A[p .. i] \leq pivot, A[i + 1 .. j - 1] > pivot, p - 1 \leq i < j$

Proof.

## Maintenance (continued):

1. Case  $A[j] \leq pivot$ : We increment  $i$  and exchange  $A[i]$  and  $A[j]$ . I'll write  $i$  for the new value and  $i_0$  for the pre-increment value,  $i = i_0 + 1$ . I'll write  $A_0[i]$  and  $A_0[j]$  for the pre-exchange array values. ( $i_0 < j$  so  $i < j + 1$ , so that part of LI holds for  $j + 1$ .)
  - ▶ We have added  $A_0[j] \leq pivot$  to the  $\leq$  range and extended its size by incrementing  $i$ , so  $A[p .. i] \leq pivot$  holds.
  - ▶ We have moved  $A_0[i_0 + 1]$ . It was either the first element of the  $>$  range, or the  $>$  range was empty and it was the first unexamined element (and the "exchange" didn't move it).
  - ▶ In either case, the  $>$  range (empty or not), moves right one step: it lost  $A[i_0 + 1] = A[i]$  and it now starts at  $A[i + 1]$  and runs to  $A[j]$ . That is,  $A[i + 1 .. (j + 1) - 1] > pivot$ , so the LI holds for  $j + 1$ .

# Partition, Proof of Correctness

$$\text{LI} : A[p .. i] \leq \text{pivot}, A[i + 1 .. j - 1] > \text{pivot}, p - 1 \leq i < j$$

## Proof.

**Termination:** After the loop ends,  $j = r$  (the loop does not cover  $r$ ), so the loop invariant gives

- ▶  $A[p .. i] \leq \text{pivot}$
- ▶  $A[i + 1 .. r - 1] > \text{pivot}$
- ▶  $p - 1 \leq i < r$

The algorithm's final step is to exchange  $A[i + 1]$  and  $A[r]$ .

This shifts the  $>$  range (empty or not) right one index (see reasoning from Maintenance case 1). So  $A[i + 2 .. r] > \text{pivot} = A[i + 1]$ .

Let  $q = i + 1$ , the return value. Then we have

- ▶  $A[p .. i - 1] \leq A[q] < A[q + 1 .. r]$
- ▶  $p \leq q \leq r$



Running time of Partition is clearly  $\Theta(n)$  in all cases.

Running time of Quick Sort:

- ▶ Best case is when the array is partitioned into two equal parts.
- ▶ In this case the recurrence is  $T(n) = 2T(n/2) + \Theta(n)$ .
- ▶ We already know this is  $\Theta(n \log n)$ .

# Running Time: Worst Case

- ▶ The worst case happens when the pivot partitions the array into two sub-arrays of size  $n-1$  and  $0$ .
- ▶ This happens when the array is already sorted.
- ▶ Thus we have:

$$\begin{aligned}T(n) &= T(n-1) + T(0) + \Theta(n) \\ &= T(n-1) + \Theta(n) \\ &= \sum_{j=0}^n \Theta(j) = \Theta\left(\frac{n(n+1)}{2}\right) = \Theta(n^2)\end{aligned}$$

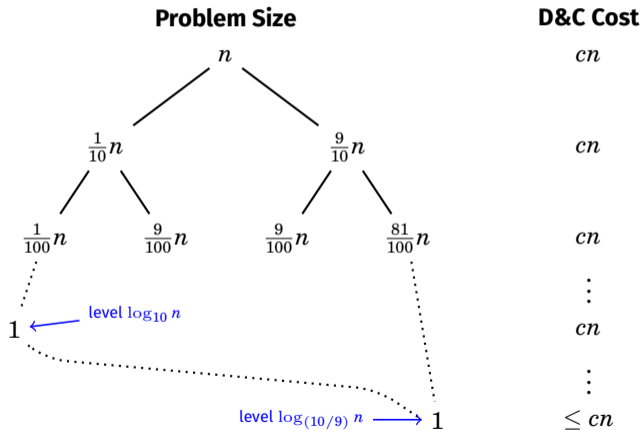


# Running Time: Average Case

- ▶ Claim: the average runtime seems to be  $O(n \log n)$ .
- ▶ This means that on average we hit a “good” case.
- ▶ This is quite atypical, as usually the average case is no better than the worst case.
- ▶ What explains Quick Sort’s luck?

# Running Time: Average Case

What happens if the pivot divides the array into two sub-arrays of  $0.9n$  and  $0.1n$ ?



# Running Time: Average Case

Analysis of Unlucky Case (0.1 – 0.9 split):

- ▶ There are  $1 + \log_{(10/9)} n$  levels and each has  $O(n)$  cost.
- ▶ The total cost is therefore  $O(n \log n)$ .

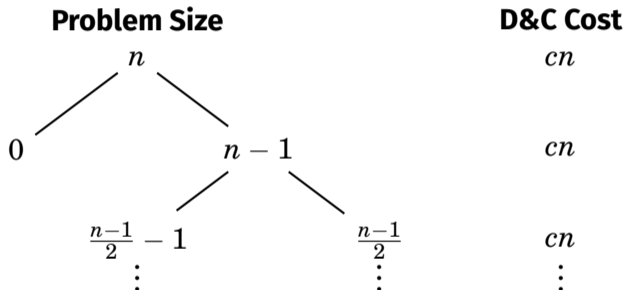
So Quick Sort is not *that* sensitive to how good the pivot is.

What about a different kind of bad luck?

- ▶ What happens if occasionally it is as bad as can be?
- ▶ Suppose every other iteration the pivot is the largest element.

# Running Time: Average Case

Suppose every other iteration the pivot is the largest element.



We simply double the number of levels, it is still  $O(n \log(n))$ .

## Probabilistic Analysis

- ▶ Remember the average runtime analysis of Insertion Sort.
- ▶ We averaged the running time over a particular **distribution** of inputs — we used a **uniform distribution**: all inputs equally likely.
- ▶ We have to know the distribution of the input — and be able to calculate an average over it!

## Randomized Analysis

- ▶ We can *change the algorithm* to introduce randomness. But it still must *definitely* behave according to its specification.
- ▶ By adding randomness, we can make the input distribution *irrelevant*, making it easier to calculate the average (or **expected**) case behavior.

# Randomized Quicksort

- ▶ We have a random number generator  $\text{Random}(p,r)$  which produces numbers between  $p$  and  $r$ , each with equal probability. In practice most random number generators produce pseudo-random numbers.
- ▶ The selected number is the pivot index.
- ▶ When analyzing the running time of a randomized algorithm we take the **expected** run time over all inputs.

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**Algorithm 3** RandomizedPartition( $A, p, r$ )

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**Ensure:** (same as Partition)

- 1:  $i \leftarrow \text{Random}(p, r)$
  - 2: exchange  $A[i] \leftrightarrow A[r]$
  - 3: **return** Partition( $A, p, r$ )
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**Algorithm 4** RandomizedQuicksort( $A, p, r$ )

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**Ensure:** (same as Quicksort)

- 1: **if**  $p < r$  **then**
  - 2:    $q \leftarrow$  RandomizedPartition( $A, p, r$ )
  - 3:   RandomizedQuicksort( $A, p, q - 1$ )
  - 4:   RandomizedQuicksort( $A, q + 1, r$ )
  - 5: **end if**
-



# Rigorous Worst Case Analysis of Quicksort

Let  $T(n)$  be the worst case running time for quicksort (or randomized quicksort). It is described by

$$T(n) \leq \max_{0 \leq q \leq n-1} (T(q) + T(n - q - 1)) + an$$

for some  $a > 0$ .

That is, the worst case happens when, on each recursive call, we pick the worst pivot, resulting in the worst (maximum) combined run times on the sub-problems.

We **guess** that  $T(n) = O(n^2)$ , and now we'll **prove** it.

# Rigorous Worst Case Analysis of Quicksort

$$T(n) \leq cn^2$$

## Proof by induction.

- ▶ Base case: We must show  $T(1) \leq c$ . Trivial.
- ▶ Inductive case: We must show  $T(n) \leq cn^2$ .
- ▶ Inductive hypothesis: Assume  $T(k) \leq ck^2$  for all  $1 \leq k < n$ .
- ▶ Calculate:

$$\begin{aligned} T(n) &\leq \max_{0 \leq q \leq n-1} (T(q) + T(n-q-1)) + an \\ &\leq c \max_{0 \leq q \leq n-1} (q^2 + (n-q-1)^2) + an \end{aligned}$$

- ▶ The expression  $(q^2 + (n-q-1)^2)$  is a convex function, achieving a maximum at the endpoints: 0 and  $n-1$ .
- ▶ In those endpoints the value is  $(n-1)^2$ .

# Rigorous Worst Case Analysis of Quicksort

## Proof by induction, Cont.

► Therefore:

$$\begin{aligned}T(n) &\leq \max_{0 \leq q \leq n-1} (T(q) + T(n - q - 1)) + an \\&\leq c \max_{0 \leq q \leq n-1} (q^2 + (n - q - 1)^2) + an \\&\leq cn^2 - c(2n - 1) + an \\&= cn^2 - (2c - a)n + c \\&\leq cn^2 - (2c - a)n + cn && \text{because } n \geq 1 \\&= cn^2 - (c - a)n\end{aligned}$$

► We must pick a large enough  $c$  so that  $c \geq a$ .



# Rigorous Worst Case Analysis of Quicksort

- ▶ We just proved an upper bound to the worst case runtime:  
 $T(n) = O(n^2)$ .
- ▶ Previously we have seen a case where the run time is quadratic.  
That is, we knew  $T(n) = \Omega(n^2)$ .
- ▶ So when  $T(n)$  represents the worst-case performance,  
 $T(n) = \Theta(n^2)$ .

# Average Case Analysis: Method 1

The average (ie, **expected**) run time for Randomized-Quicksort on an array of size  $n$  is described by the following equation:

$$\begin{aligned}T(n) &= \frac{1}{n} \sum_{q=0}^{n-1} (T(q) + T(n - q - 1)) + cn + \Theta(1) \\ &= \frac{2}{n} \sum_{q=0}^{n-1} T(q) + cn + \Theta(1)\end{aligned}$$

- ▶ We wrote  $cn + \Theta(1)$  rather than  $\Theta(n)$  since we can assume we do “everything” every time we call Partition.
- ▶ This is a worst case assumption that allows us to do something really nice mathematically.

# Average Case Analysis: Method 1

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

$$nT(n) = 2 \sum_{q=0}^{n-1} T(q) + cn^2 + \Theta(n) \quad \text{multiply by } n$$

$$(n+1)T(n+1) = 2 \sum_{q=0}^n T(q) + c(n+1)^2 + \Theta(n)$$

multiply by  $n+1$

$$(n+1)T(n+1) - nT(n) = 2T(n) + \Theta(n) \quad \text{subtract}$$

$$(n+1)T(n+1) = (n+2)T(n) + \Theta(n) \quad \text{simplify}$$

# Average Case Analysis: Method 1

- ▶ Starting from:  $(n + 1)T(n + 1) = (n + 2)T(n) + \Theta(n)$
- ▶ Divide by  $(n + 1)(n + 2)$  to get:  $\frac{T(n+1)}{n+2} = \frac{T(n)}{n+1} + \Theta\left(\frac{1}{n}\right)$
- ▶ Define  $g(n) = \frac{T(n)}{(n+1)}$
- ▶ So:  $g(n + 1) = g(n) + \Theta\left(\frac{1}{n}\right)$
- ▶ Then:  $g(n) = \Theta\left(\sum_{k=1}^{n-1} \frac{1}{k}\right) = \Theta(\log n)$
- ▶ Going back:  $T(n) = (n + 1)g(n) = \Theta(n \log n)$

## Average Case Analysis: Method 2

- ▶ The total cost is the sum of the costs of all the calls to RandomizedPartition.
- ▶ The cost of a call to RandomizedPartition is  $O(\#\mathbf{for}$  loop executions), which is  $O(\#\text{comparisons})$ .
- ▶ The expected cost of RandomizedQuicksort is  $O(\text{expected } \#\text{comparisons})$ .
- ▶ Notice that once a key  $x_k$  is chosen as pivot, the elements to its left will never be compared to the elements to its right.



## Average Case Analysis: Method 2

- ▶ Consider  $\{x_i, x_{i+1}, \dots, x_{j-1}, x_j\}$ , the set of keys in sorted order.
- ▶ Any two keys here are compared only if one of them is pivot and that is the last time they are all in the same partition.
- ▶ Each key is equally likely to be chosen as the pivot.
- ▶  $x_i$  and  $x_j$  can be compared only if one of them is pivot and this will only happen if this is the first pivot from the set  $\{x_i, x_{i+1}, \dots, x_{j-1}, x_j\}$ .
- ▶ The probability of this is  $\frac{2}{(j-i+1)}$ .

## Average Case Analysis: Method 2

The expected number of comparisons is:

$$\begin{aligned}\sum_{i < j} \frac{2}{j - i + 1} &= \sum_{i=1}^{n-1} \sum_{j=i+1}^n \frac{2}{j - i + 1} \\ &= \sum_{i=1}^{n-1} \sum_{k=1}^{n-i} \frac{2}{k + 1} \\ &\leq \sum_{i=1}^{n-1} \sum_{k=1}^n \frac{2}{k} \\ &= 2(n - 1)H_n = O(n \log n)\end{aligned}$$

where  $H_n$  is the  $n$ th Harmonic number (see A.7 in the Appendix)