

Today:

- Quicksort
(or *Randomized* algorithm for sorting)

Yet another sorting algorithm, but:

- Sorts “in place” (doesn’t require additional array)
 - like insertion sort
 - unlike merge sort
- Very practical (with tuning)

Quicksort — Divide-and-conquer algorithm

1. **Divide:** Partition array into 2 subarrays such that elements in lower part \leq elements in higher part.
2. **Conquer:** Recursively sort 2 subarrays.
3. **Combine:** Trivial (because in place).

Key:

Linear-time ($\Theta(n)$) partitioning procedure

$$\left| \leq x \quad |x| \quad \geq x \right|$$

PARTITION procedure

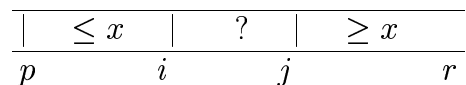
```

PARTITION( $A, p, r$ )
  (Partition  $A[p..r]$  around random
  element  $x = A[k]$ )
   $k \leftarrow \text{Random}(p \dots r)$ 
   $x \leftarrow A[k]$ 
   $i \leftarrow p - 1$ 
   $j \leftarrow r + 1$ 

  while TRUE do
    repeat  $j \leftarrow j - 1$  until  $A[j] \leq x$ 
    repeat  $i \leftarrow i + 1$  until  $A[i] \geq x$ 
    if  $i < j$  then exchange  $A[i] \leftrightarrow A[j]$ 
    else quit (and return  $j$ )
  
```

Correctness proof idea

Loop invariant:



Time = $\Theta(n)$ for n -element subarray

Quicksort — Recursive algorithm

```

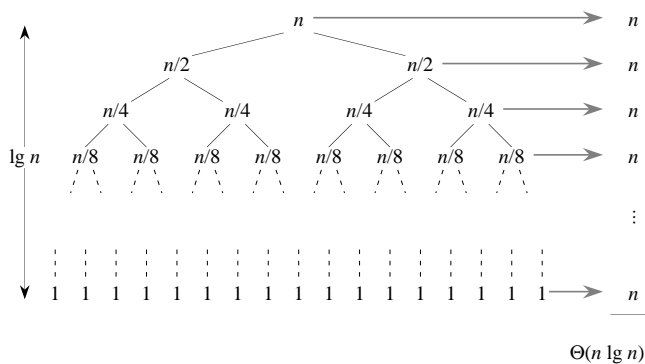
QUICKSORT( $A, p, r$ )
  if  $p < r$ 
    then  $q \leftarrow \text{PARTITION}(A, p, r)$   $\triangleright$  around  $A[r]$ 
         QUICKSORT( $A, p, q - 1$ )
         QUICKSORT( $A, q + 1, r$ )
  
```

Initial call: QUICKSORT($A, 1, \text{length}[A]$)

Analysis of QUICKSORT (best case)

If we're lucky, PARTITION always splits array evenly

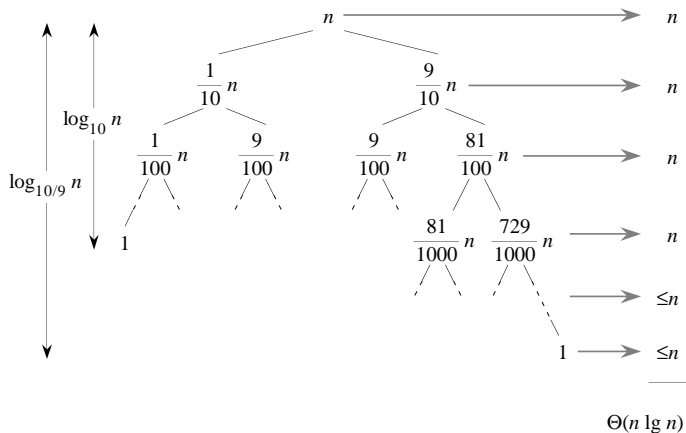
$$T(n) = 2T(n/2) + \Theta(n)$$



Suppose the split is $\frac{1}{10} : \frac{9}{10}$

$$T(n) = T(n/10) + T(9n/10) + \Theta(n)$$

$$= \Theta(n \lg n) \quad \text{Still lucky!}$$



Analysis of QUICKSORT (worst case)

How might we be unlucky?

- one side of partition has 1 element

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

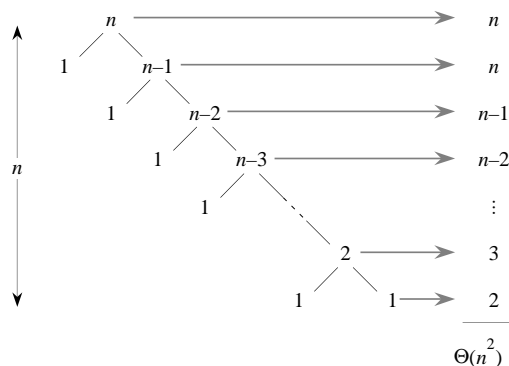
$$= T(n-1) + \Theta(n) \quad \text{because } T(1) = \Theta(1)$$

$$= \sum_{k=1}^n \Theta(k) = \Theta\left(\sum_{k=1}^n k\right)$$

$$= \Theta(n^2) \quad \text{[arithmetic series]}$$

Recursion Tree

- Left nodes ($\Theta(1)$) add up to $\Theta(n)$
- Root and right nodes add up to $\Theta\left(\sum_{k=1}^n k\right) = \Theta(n^2)$
- Total: $\Theta(n) + \Theta(n^2) = \Theta(n^2)$



- Our “random” elements x are $1, 2, 3, 4 \dots n$
- Does not look very random !

- Randomized algorithm - running time varies even for the same input. Therefore, we can say the running time is “this much” on the *average*, or it is “this much” with *certain probability*.
- Assume random number generates independent choices throughout algorithm.
- For technical reasons, assume all input elements are distinct. Picking index of element is equivalent to picking element.

Key observations

- the running time bounded by $O(n)$ times the depth of the recursion tree (as seen on earlier pictures)
- “nice splits” happen with “nice probability”
- if we have large enough number of trials, and each trial has “nice probability” of success, we have many successes with high probability

Lucky partitions

Let a_i be the i -th smallest element in $A[\cdot]$ (i.e., with the rank i).

Let D_i denote the depth of a_i in the recursion tree. The total tree depth is $D = \max_i D_i$.

We say a split (say, of $A[p \dots r]$) is “lucky” for a_i , if the part where a_i ends up in is of size $\leq 3/4(r - p + 1)$, i.e., if we reduce the array size by a factor of $3/4$.

After $\log_{4/3} n$ lucky splits, a_i is a leaf !

“Lucky” lemma

Lemma: At any time, a_i is lucky with probability at least $1/4$ (could be even higher).

Proof: If $i \leq n/2$, choosing x to be any element with rank in $\{n/2 \dots 3n/4\}$ creates a lucky split. If $i \geq n/2$, situation is symmetric.

The analysis

Let $l = C \log_{4/3} n$, where C is a “large” constant (set later).

What is the probability that $D_i > l$? It is at most the probability that in l trials, each having success probability $\geq 1/4$, we were *not* successful $\geq l - t = l - \log_{4/3} n$ times (since $\log_{4/3} n$ lucky splits are enough to make a_i a leaf).

The latter probability is at most

$$\binom{l}{l-t} (1 - 1/4)^{l-t}$$

(will elaborate on that later)

The analysis ctd.

$$\begin{aligned} \binom{l}{l-t} (1 - 1/4)^{l-t} &= \binom{l}{t} (3/4)^{l-t} \\ &\leq \left(\frac{el}{t}\right)^t (3/4)^{l-t} \\ &= \left(\frac{eC \log_{4/3} n}{\log_{4/3} n}\right)^t (3/4)^{(C-1) \log_{4/3} n} \\ &= \frac{(eC)^{\log_{4/3} n}}{n^{C-1}} \\ &= n^{\log_{4/3} eC - (C-1)} \end{aligned}$$

When C is large enough (e.g., 100), the probability is smaller than $1/n^2$.

Finishing the analysis

We proved that $\Pr[D_i > l] \leq 1/n^2$.

Therefore, the probability that *there exists* i such that $D_i > l$ is at most

$$\Pr[D_1 > l] + \Pr[D_2 > l] \dots \Pr[D_n > l] \leq n \cdot 1/n^2 = 1/n$$

Therefore probability that $\max_i D_i \leq l$ is at least $1 - 1/n$.

Appendix

Why is the probability that “in l trials, each having success probability $\geq 1/4$, we were *not* successful $\geq l - t$ times” at most

$$\binom{l}{l-t} (3/4)^{l-t}$$

?

- if we were not successful $\geq l - t$ times, we were not successful during some set of $l - t$ trials
- the probability we were not successful during a *fixed* set of exactly $l - t$ trials is at most $(3/4)^{l-t}$
- there are $\binom{l}{l-t}$ subsets of trials of size $l - t$