

CSc445 Algorithms

Everything you always wanted to know about Quick Sort,

What lessons could QuickSort teaches us about other algorithms

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Based on slides curacy of
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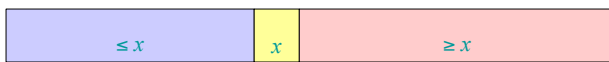
QuickSort – example of the divide-and-concourse paradigm

- Proposed by C.A.R. Hoare in 1962.
- Sorts “in place” (no need for extra space). Like insertion sort, but not like merge sort.
- Very practical (with tuning).

Divide and conquer

Quicksort an n -element array:

1. **Divide:** Partition the array into two subarrays around a **pivot** x such that elements in lower subarray $\leq x$ elements in upper subarray.
2. **Conquer:** Recursively sort the two subarrays.
- **Combine:** Trivial.



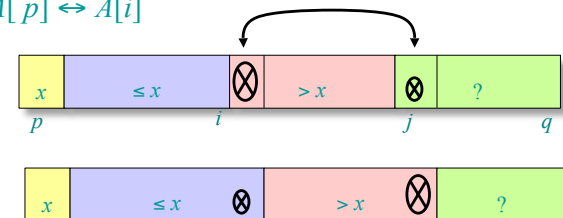
Key: Linear-time partitioning subroutine.

Partitioning subroutine

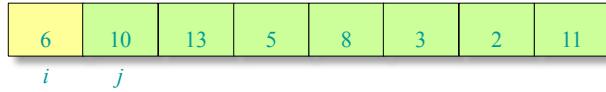
```
PARTITION( $A, p, q$ )  $\triangleright A[p..q]$ 
 $x \leftarrow A[p]$   $\triangleright$  pivot =  $A[p]$ 
 $i \leftarrow p$ 
for  $j \leftarrow p + 1$  to  $q$   $\triangleright$   $j$  is hunting for small keys
  do if  $A[j] \leq x$   $\triangleright$  Should send  $A[j]$  to the left.
  then{
     $i \leftarrow i + 1$   $\triangleright$  Now  $A[i] > x$ 
    exchange  $A[i] \leftrightarrow A[j]$   $\triangleright$  Fix  $A[i] > x$ 
  }
exchange  $A[p] \leftrightarrow A[i]$ 
return  $i$ 
```

Running time = $O(n)$
for n elements.

Invariant:

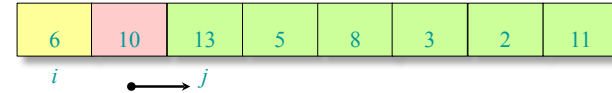


Example of partitioning



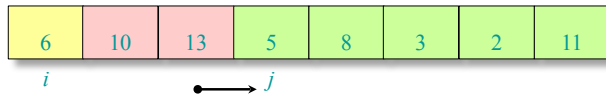
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Example of partitioning

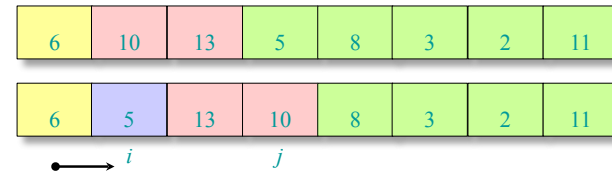


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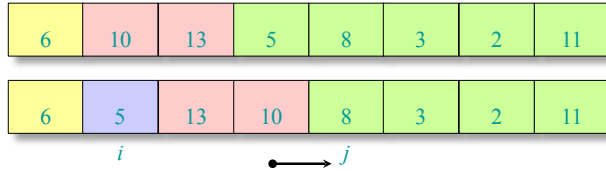
Example of partitioning



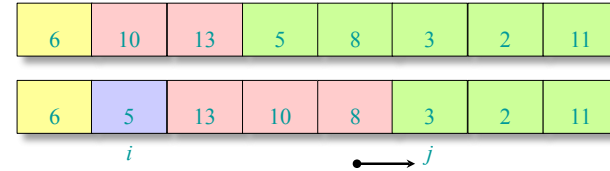
Example of partitioning



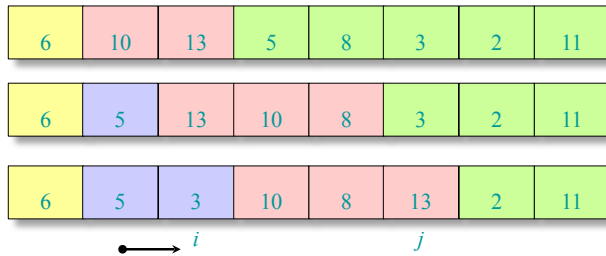
Example of partitioning



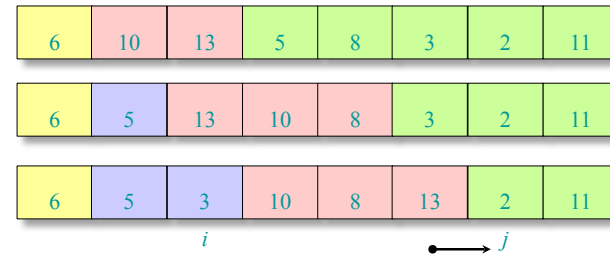
Example of partitioning



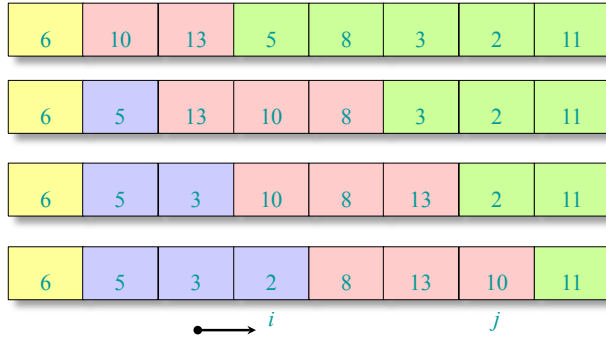
Example of partitioning



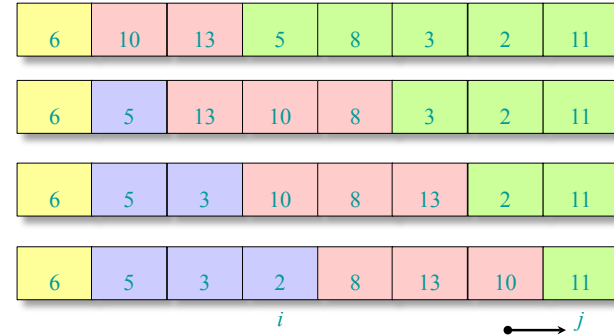
Example of partitioning



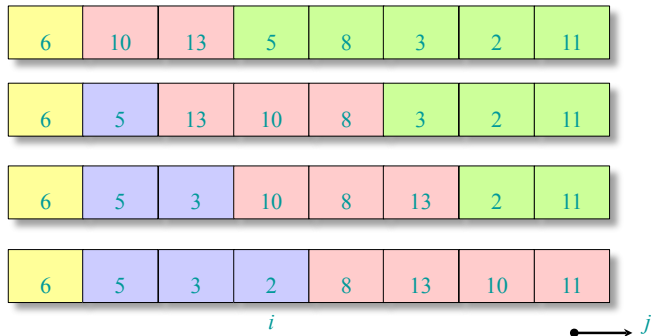
Example of partitioning



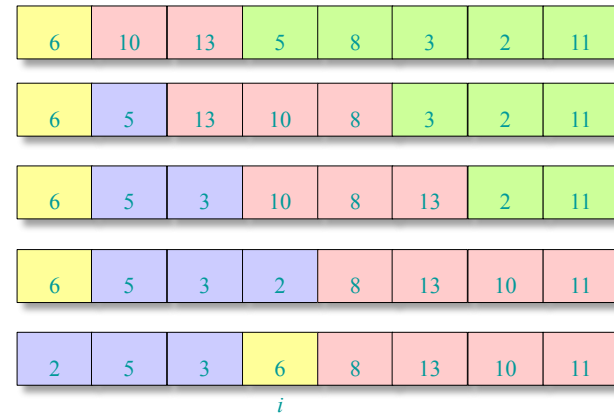
Example of partitioning



Example of partitioning



Example of partitioning



Pseudocode for quicksort

```
QUICKSORT( $A, p, r$ )  
  if  $p < r$  //do something only if contains at least 2 keys  
    then  $q \leftarrow$  PARTITION( $A, p, r$ ) //both perform partition, and  
        return index of pivot  
        QUICKSORT( $A, p, q-1$ ) //QS left part  
        QUICKSORT( $A, q+1, r$ ) //QS right part
```

Initial call: AUICKSORT($A, 1, n$)

Analysis of quicksort

- Assume all input elements are distinct.
- In practice, there are better partitioning algorithms for when duplicate input elements may exist.
- Let $T(n)$ = worst-case running time on an array of n elements.

Worst-case of quicksort

- Input sorted or reverse sorted.
- Partition around min or max element.
- One side of partition always has no elements.

$$\begin{aligned}T(n) &= T(0) + T(n-1) + \Theta(n) \\ &= \Theta(1) + T(n-1) + \Theta(n) \\ &= T(n-1) + \Theta(n) \\ &= \Theta(n^2) \quad (\text{arithmetic series})\end{aligned}$$

Worst-case recursion tree

$$T(n) = T(0) + T(n-1) + cn$$

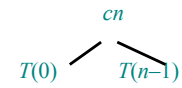
Worst-case recursion tree

$$T(n) = T(0) + T(n-1) + cn$$

$T(n)$

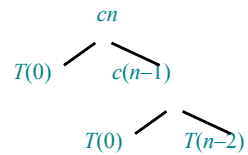
Worst-case recursion tree

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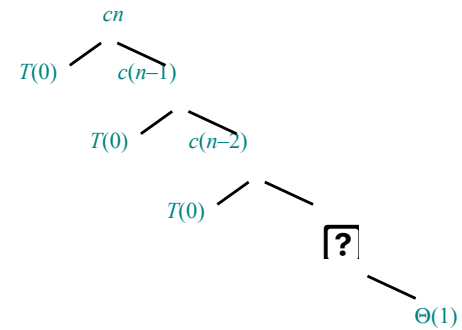
Worst-case recursion tree

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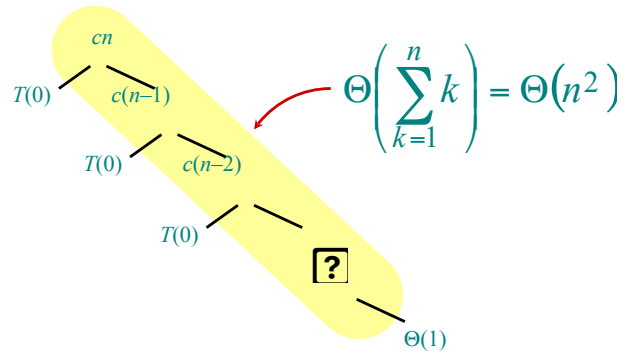
Worst-case recursion tree

$$T(n) = T(0) + T(n-1) + cn$$



Worst-case recursion tree

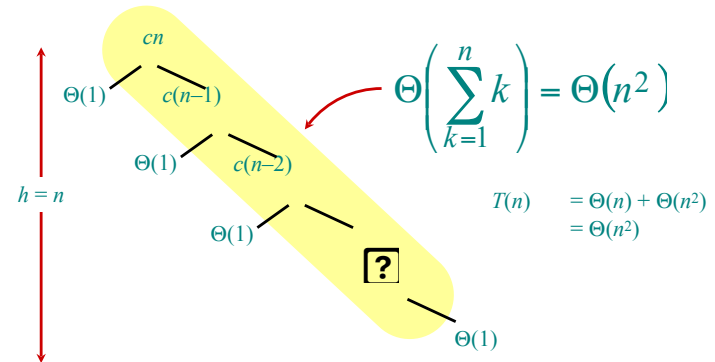
$$T(n) = T(0) + T(n-1) + cn$$



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Worst-case recursion tree

$$T(n) = T(0) + T(n-1) + cn$$



Best-case and almost best-case analysis

If we are lucky, PARTITION splits the array evenly:

$$\begin{aligned} T(n) &= 2T(n/2) + \Theta(n) \\ &= \Theta(n \lg n) \end{aligned} \quad \text{(same as merge sort)}$$

What if the split is $\frac{1}{10} : \frac{9}{10}$?

That is, both sub-arrays contains at least 10% of the keys (possibly more)

$$T(n) = T\left(\frac{1}{10}n\right) + T\left(\frac{9}{10}n\right) + \Theta(n)$$

We call such a partition an **almost-optimal** partition.

What is the running time in this case?

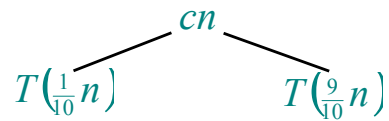
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Analysis of “almost-best” case

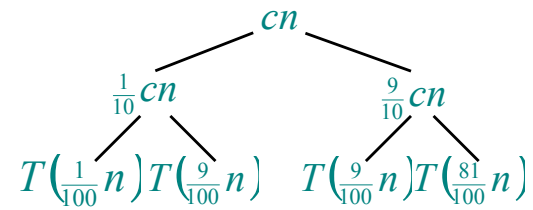
$$T(n)$$

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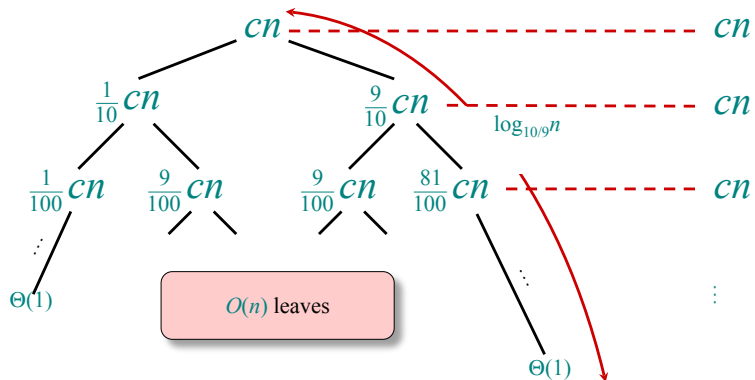
Analysis of “almost-best” case



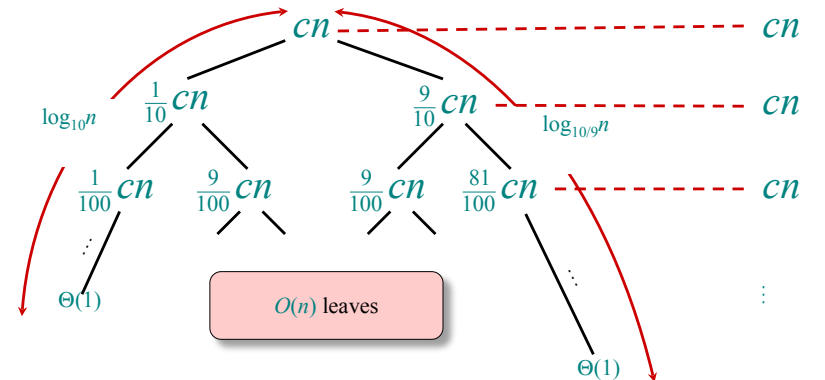
Analysis of “almost-best” case



Analysis of “almost-best” case



Analysis of “almost-best” case



$$cn \log_{10} n \leq$$

$$T(n) \leq cn \log_{10} n + O(n) \leq 8cn \log_2 n$$

QS needs $O(n \log n)$ if partition are almost optimal

Each time the algorithm invested some work, it moves a key from one location to another

Consider a key x .

When the algorithm starts, it is in an array of size n

Then x is shifted into an array of size. $\leq (0.9) \cdot n$

Next, x “ “ “ “ of size $\leq (0.9)^2 \cdot n$

Next, x “ “ “ “ of size. $\leq (0.9)^3 \cdot n$

⋮

After k times that x was shifted, its array's size $\leq (0.9)^k \cdot n$

Max time that x is shifted:

$$(0.9)^k n \leq 1 \quad \text{OR} \quad k \leq \log_{(10/9)} n \leq 8 \log_2 n = O(\log n)$$

Next we need to multiply this number of the number of keys, yielding $O(n \log n)$

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Randomized quicksort

How can find a pivot that guarantees partitions with good ratios for $A[1..n]$, ?

We say that q is a **good pivot** for if

- at least 10% of the elements of $A[1..n]$ are smaller than q , and
- at least 10% of the elements of $A[1..n]$ are larger than q .



Best pivot: Pick the **median** of $A[1..n]$, as pivot.

(median – an element that is larger than half of the keys)

Then the time would obey $T(n) = cn + 2T(n/2)$

Problem – need to work too hard to find the median (best pivot), so we will do with (only) a good pivot. (of course, we could first sort :-).)

Finding a good pivot for $A[1..n]$

5-random-elements method. :

- Pick the **indices** of 5 elements at random from $A[1..n]$,
- For $k=1$ to 5

$$X[k] = A[\lfloor n \cdot \text{rand}() \rfloor]$$

$A[1..n]$



- Set q to be the median of $X[1..5]$

Finding a good pivot for $A[1..n]$

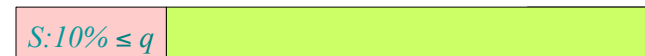
5-random-elements method. : Pick 5 elements at random from $A[1..n]$, and set q to be their median.

What is the probability that q is **not** a good pivot ?

- Let S be the elements of $A[1..n]$ which are the 10% smallest.
- The probability that an elements picked at random is in S is 0.1 .
- q is in S only if **at least 3** of the 5 elements that we pick are in S .
- The probability that this happens is

$$= 0.1^5 + 5 \cdot 0.1^4 \cdot 0.9 + 10 \cdot 0.1^3 \cdot 0.9^2 = 0.00001 + 0.00045 + 0.00810 = 0.00856$$

- This is also the probability that q is in the 10% largest elements.
- In other words: with probability $\geq 98\%$, q is a good pivot.



Putting it together

- If we performed a partition which is **not** almost optimal, nothing dramatically bad happens, we just wasted some time. Each such partition takes linear time, but has no effect.
- However, each partition is, with probability $\geq 98\%$ is good, and we obtain an almost-optimal pivot.
- Hence the expected time of QuickSort (if the 5 random keys methods is used) is

$$O(n \log n) + 0.02 \cdot O(n \log n) = O(n \log n)$$

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Randomized quicksort – cont Finding good pivots

Putting it together, during QS, each time that we need to find a pivot, we use the “5 random elements” method.

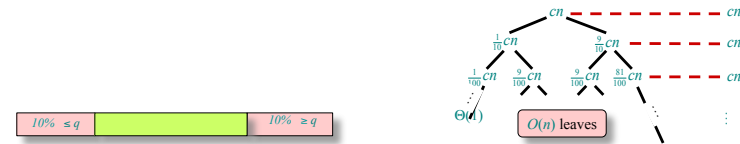
With probability 98%, we find a good pivot.

The overall time that we spend on good partitions is much smaller than the time we spent on bad partitions.

(note – bad partitions are not harmful – they are just not helpful)

So the recursions formula $T(n) = cn + T(n/10) + T(n 9/10)$ still apply, leading to running time $O(n \log n)$.

This is expected running time – there is a chance that the actual running time is $\Theta(n^2)$, but the probability that it happens is very slim.



Quicksort in practice

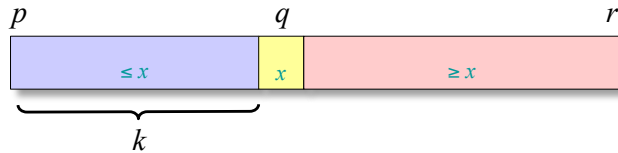
- Quicksort is a great general-purpose sorting algorithm.
- Quicksort is typically over twice as fast as merge sort.
- Quicksort behaves well even with caching and virtual memory.

Median Selection

- (CLRS Section 9.2, page 185).
- For $A[1..n]$ (all different elements) we say that the rank of x is i if exactly $i-1$ elements in A are smaller than x .
- In particular, the median is the $\lfloor n/2 \rfloor$ -smallest.
- To find the median, we could sort and pick $A[\lfloor n/2 \rfloor]$ (taken $O(n \log n)$).
- We can do better.

Median Selection-cont

```
RS(A, p, r, i){  
  //Randomize Selection: Returns i't smallest element in A[p..r].  
  //Assumption: Input is valid and elements are different.  
  •If  $p==r$  return  $A[p]$   
  • $q=\text{PARTITION}(A,p,r)$  ;  
    •//Partition using the 5-random element method  
  • $k=q-p$   
  •If  $i==k+1$  return  $A[q]$   
  •If  $i<k$  return  $\text{RS}(A, p, q-1, i)$  // Note the difference from QS  
  •Else return  $\text{RS}(A, q+1, r, i-k-1)$   
}
```



Time analysis

- Recall: With high probability, we pick a good pivot:
 - Not in the 10% smallest or largest:
- Hence, we get rid of at least 10% of the elements of A
- So, $T(n) = cn + T(0.9n)$.
 - $T(n) = c(n + 0.9n + 0.9^2n + 0.9^3n + \dots) =$
 $cn(1 + 0.9 + 0.9^2 + 0.9^3 + \dots) =$
 $cn(1/(1-0.9)) = O(n)$.
- So the expected time is linear. (yuppie)

As in the case of QS, partitions which are not good are not harmful, just not helpful.