# Video 15: Runtime of Quicksort COMS10017 - (Object-Oriented Programming and) Algorithms

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```
Require: array A of length n

if n \le 1 then

return A

else

i \leftarrow Partition(A)

QUICKSORT(A[0, i - 1])

QUICKSORT(A[i + 1, n - 1])
```

Algorithm QUICKSORT

Require: array A of length n if  $n \le 1$  then return A else  $i \leftarrow Partition(A)$ QUICKSORT(A[0, i - 1])QUICKSORT(A[i + 1, n - 1])

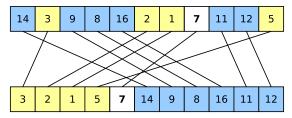
Algorithm QUICKSORT

14	3	9	8	16	2	1	7	11	12	5	
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14	3	9	8	16	2	1	7	11	12	5	
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1	2	3	5	7	8	9	11	12	14	16
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# Runtime of Quicksort

**Runtime:** T(n): worst-case runtime on input of length n

$$T(1) = O(1) \quad (\text{termination condition})$$
  
$$T(n) = O(n) + T(n_1) + T(n_2) ,$$

where  $n_1, n_2$  are the lengths of the two resulting subproblems.

**Observe:**  $n_1 + n_2 = n - 1$ 

#### Worst-case:

Suppose that pivot is always the largest element

• Then, 
$$n_1 = n - 1$$
,  $n_2 = 0$ 

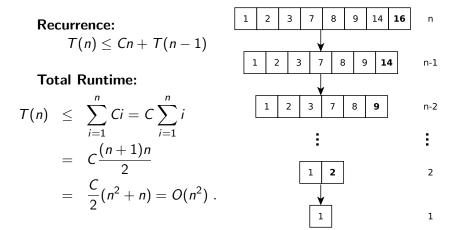
#### Best-case:

• Suppose pivot splits array evenly, i.e., pivot is the median

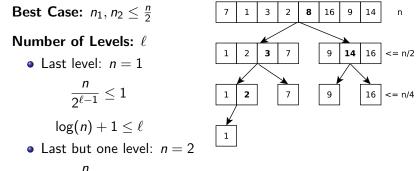
• Then, 
$$n_1 = \lfloor \frac{n-1}{2} \rfloor$$
,  $n_2 = \lceil \frac{n-1}{2} \rceil$ 

### Quicksort: Worst case

**Partition:** Let C be such that Partition() runs in time at most Cn



### Quicksort: Best case



$$\frac{n}{2^{\ell-2}} > 1$$
 which implies  $\log(n) + 2 > \ell$ 

• Hence, there are  $\ell = \lceil \log(n) \rceil + 1$  levels

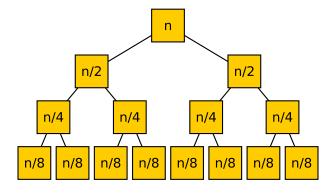
#### **Total Runtime:**

- Observe: Total runtime of Partition() in a level: O(n)
- Total runtime:  $\ell \cdot O(n) = O(n \log n)$  .

#### Good versus Bad Splits:

- It is crucial that subproblems are *roughly* balanced
- In fact, enough if  $n_1 = \frac{1}{1000}n$  and  $n_2 = n 1 n_1$  to get a runtime of  $O(n \log n)$
- Even enough if subproblems roughly balanced most of the time
- In practice, this happens most of the time, QUICKSORT is therefore usually very fast

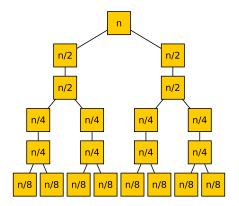
# Good versus Bad Splits: Intuition and Rough Analysis



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**Only good splits:** Recursion tree depth  $\lceil \log n \rceil + 1$ 

# Good versus Bad Splits: Intuition and Rough Analysis



**Good & bad splits alternate:** Recursion tree depth  $2 \cdot (\lceil \log n \rceil + 1)$ 

Ideal Pivot: Median

#### **Pivot Selection**

- To obtain runtime of  $O(n \log n)$ , we can spend O(n) time to select a good pivot
- There are O(n) time algorithms for finding the median
- They are complicated and not efficient in practice
- However, using such an algorithm gives  $O(n \log n)$  worst case runtime!

Idea that works in Practice: Select Pivot at random! (Implementation: exchange A[n-1] with a uniform random element A[i])

### **Randomized Algorithm**

- Randomized pivot selection turns Quicksort into a *Randomized Algorithm*
- Worst-case runtime: still  $O(n^2)$  (we may be unlucky!)
- *Expected runtime*: Since we introduce randomness, the runtime of the algorithm becomes a random variable

**Definition** (Bad Split) A split is *bad* if  $\min\{n_1, n_2\} \le \frac{1}{10}n$ .

If we select the pivot randomly, how likely is it to have a bad split?

#### Probability of a Bad Split

- Bad split if element chosen as pivot is either among smallest 0.1 fraction of elements or among largest 0.1 fraction
- Since our choice is random, this happens with probability 0.2
- Hence, in average only 1 out of 5 splits are bad
- Hence, 4 out of 5 times the algorithm makes enough progress

**Random Pivot Selection:** QUICKSORT runs in expected time  $O(n \log n)$  if the pivot is chosen uniformly at random