

# Analysis of Quicksort

Quicksort, like merge sort, applies the divide-and-conquer paradigm introduced in Section 2.3.1. Here is 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.

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 entire array  $A$ , the initial call is QUICKSORT( $A, 1, A.length$ ).

### Partitioning the array

The key to the algorithm is the PARTITION procedure, which rearranges the subarray  $A[p \dots r]$  in place.

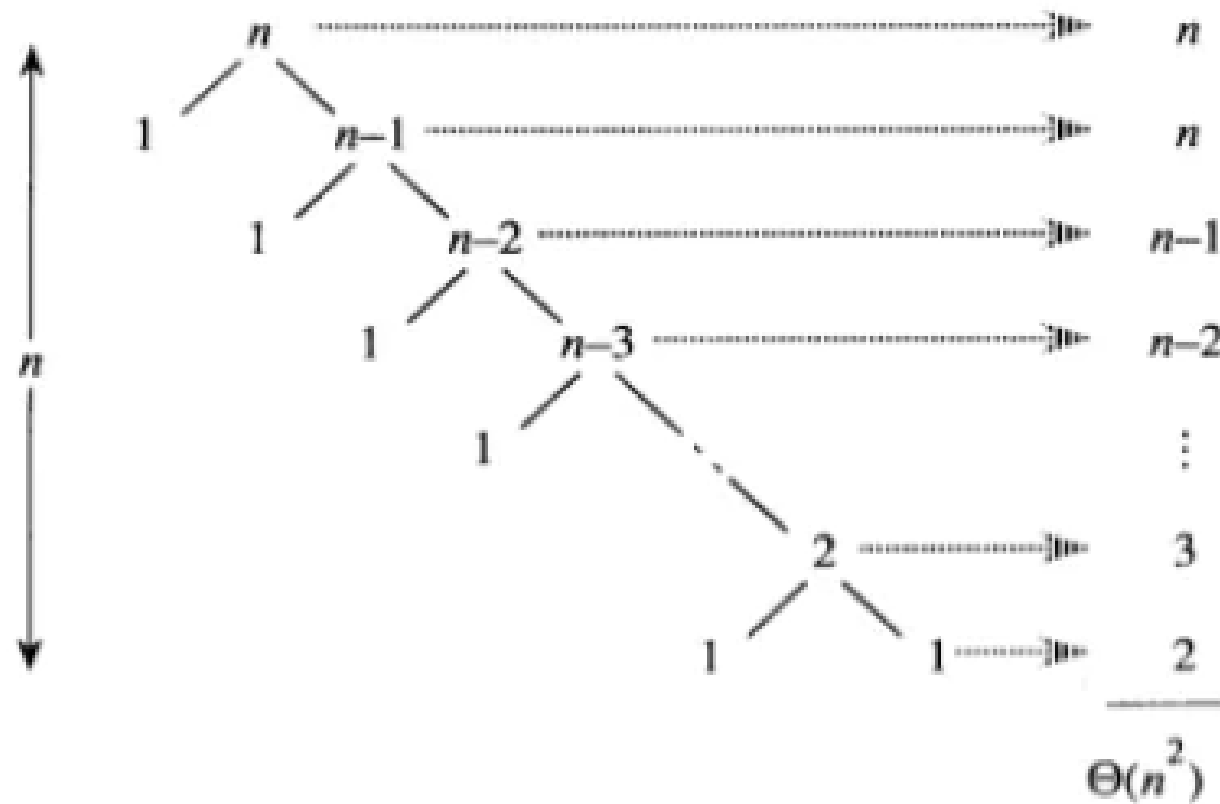
PARTITION( $A, p, r$ )

```
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  $i + 1$ 
```

# Worst Case Partitioning

- The running time of quicksort depends on whether the **partitioning** is **balanced** or not.
- ∇  $\Theta(n)$  time to partition an array of  $n$  elements
- Let  $T(n)$  be the time needed to sort  $n$  elements
- $T(0) = T(1) = c$ , where  $c$  is a constant
- When  $n > 1$ ,
  - $T(n) = T(|\text{left}|) + T(|\text{right}|) + \Theta(n)$
- $T(n)$  is maximum (**worst-case**) when either  $|\text{left}| = 0$  or  $|\text{right}| = 0$  following each partitioning

# Worst Case Partitioning



# Worst Case Partitioning

- **Worst-Case** Performance (**unbalanced**):

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

- partitioning takes  $\Theta(n)$

- $= [2 + 3 + 4 + \dots + n-1 + n] + n =$

- $= [\sum_{k=2 \text{ to } n} k] + n = \Theta(n^2)$

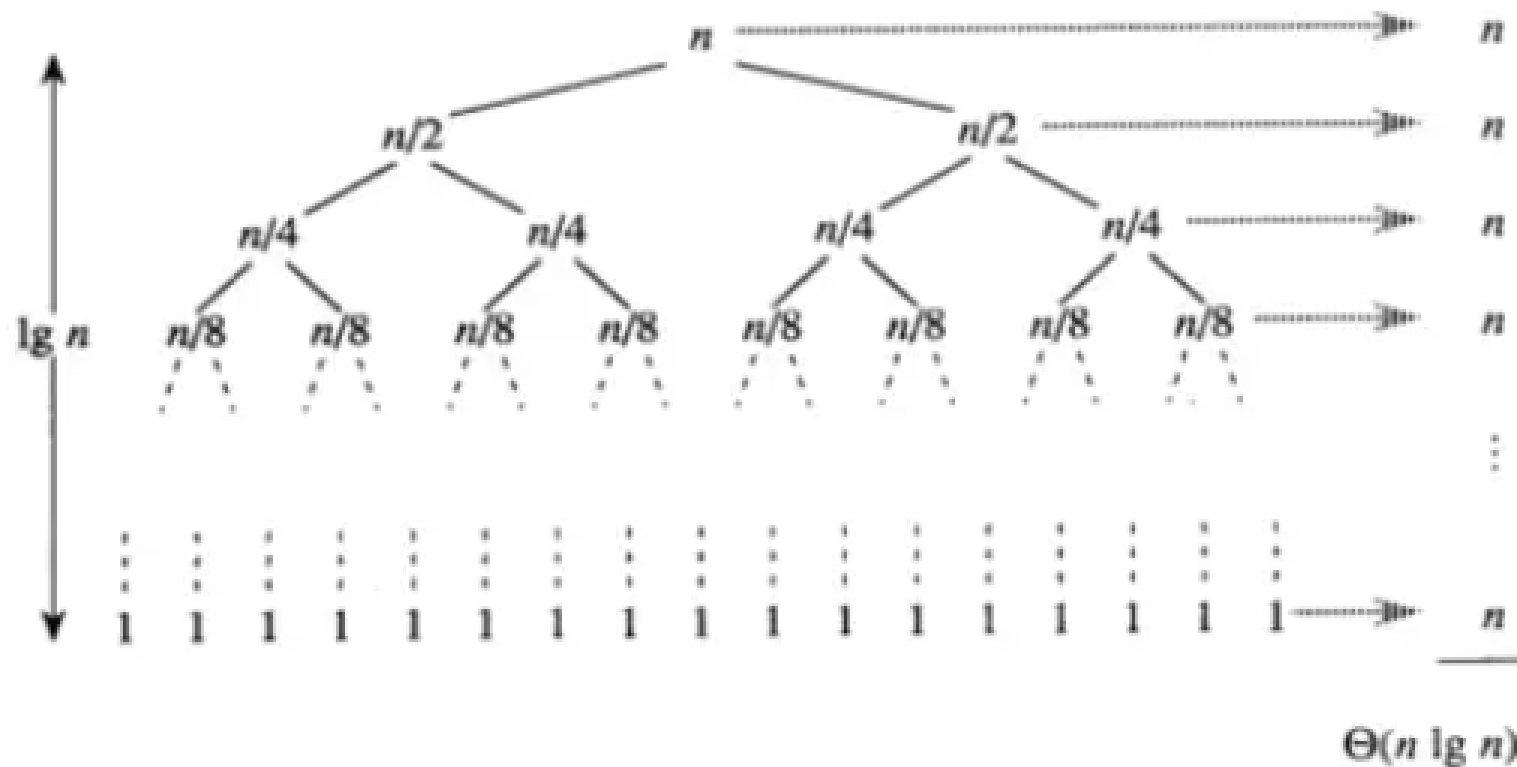
$$\sum_{k=1}^n k = 1 + 2 + \dots + n = n(n+1)/2 = \Theta(n^2)$$

- This occurs when
  - the input is **completely sorted**
- or when
  - the pivot is always the **smallest (largest)** element

# Best Case Partition

- When the partitioning procedure produces two regions of size  $n/2$ , we get the a **balanced** partition with **best case** performance:
  - $T(n) = 2T(n/2) + \Theta(n) = \Theta(n \lg n)$
- **Average** complexity is also  $\Theta(n \lg n)$

# Best Case Partitioning



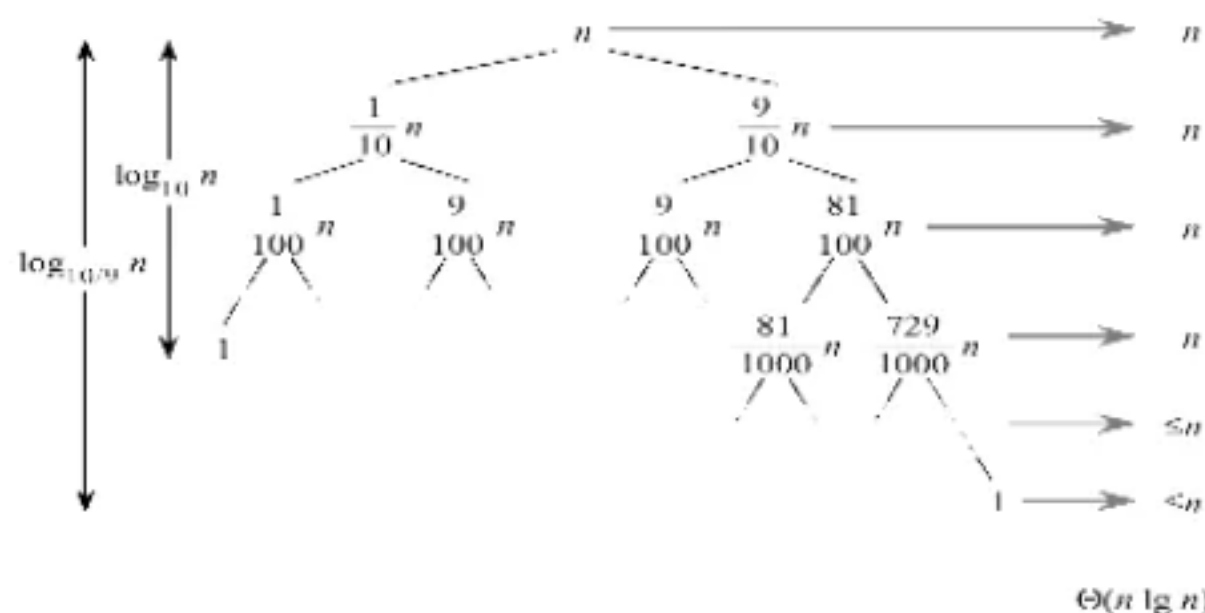


# Average Case

- Assuming **random input**, average-case running time is much closer to  $\Theta(n \lg n)$  than  $\Theta(n^2)$
- First, a more intuitive explanation/example:
  - Suppose that **partition()** always produces a **9-to-1 proportional split**. This looks quite unbalanced!
  - The recurrence is thus:
$$T(n) = T(9n/10) + T(n/10) + \Theta(n) = \Theta(n \lg n)?$$

# Average Case

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



$$\log_2 n = \log_{10} n / \log_{10} 2$$

# Average Case

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

# Average Case

- What happens if we **bad-split root node**, then **good-split** the resulting size  $(n-1)$  node?
  - We end up with **three** subarrays, size
    - $1, (n-1)/2, (n-1)/2$
  - Combined **cost of splits**  $= n + n-1 = 2n-1 = \Theta(n)$

