

Quick Sort

□ Characteristics

- sorts almost in "place," i.e., does not require an additional array
- very practical, average sort performance $O(n \log n)$ (with small constant factors), but worst case $O(n^2)$

Quick Sort – the Principle

- To understand quick-sort, let's look at a high-level description of the algorithm
- A divide-and-conquer algorithm
 - **Divide**: partition array into 2 subarrays such that elements in the lower part \leq elements in the higher part
 - **Conquer**: recursively sort the 2 subarrays
 - **Combine**: trivial since sorting is done in place

Partitioning

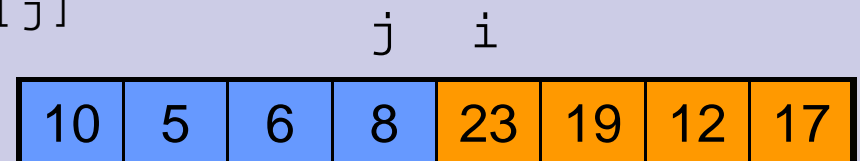
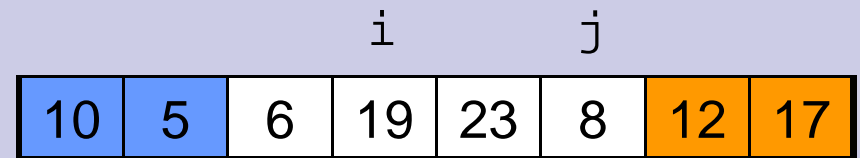
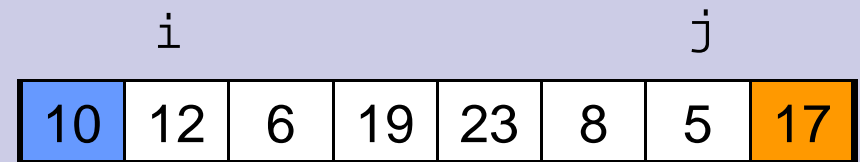
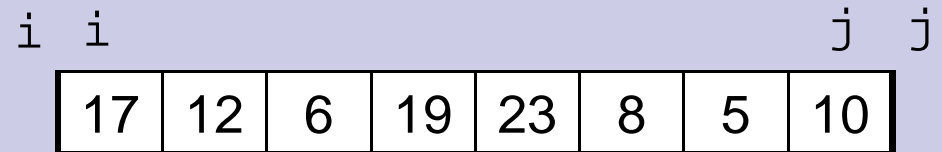
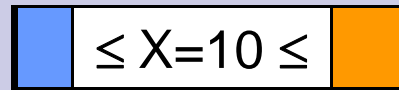
□ Linear time partitioning procedure

Partition(A, p, r)

```

01 x ← A[r]
02 i ← p - 1
03 j ← r + 1
04 while TRUE
05     repeat j ← j - 1
06         until A[j] ≤ x
07     repeat i ← i + 1
08         until A[i] ≥ x
09     if i < j
10         then exchange A[i] ↔ A[j]
11         else return j

```



Quick Sort Algorithm

- Initial call **Quicksort(A, 1, length[A])**

```
Quicksort(A, p, r)
```

```
01 if p < r
```

```
02     then q ← Partition(A, p, r)
```

```
03         Quicksort(A, p, q)
```

```
04         Quicksort(A, q+1, r)
```

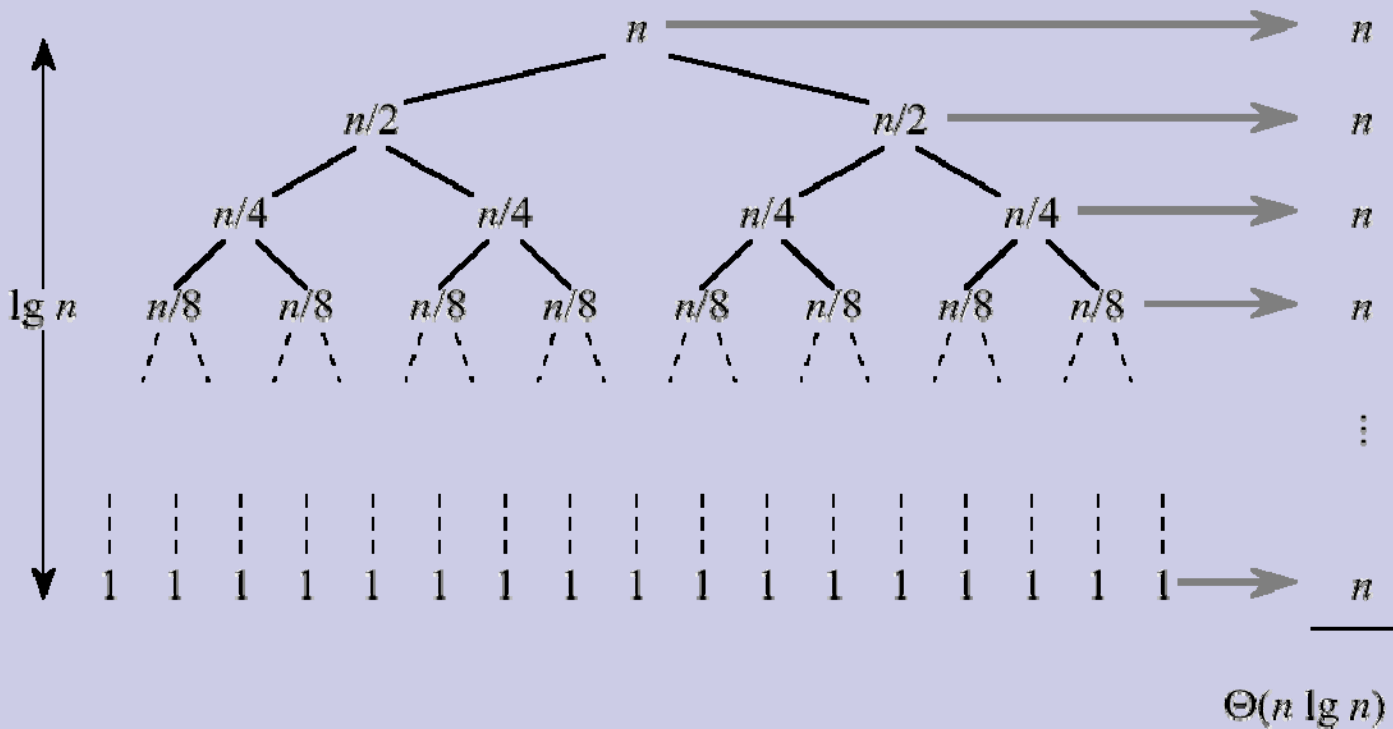
Analysis of Quicksort

- Assume that all input elements are distinct
- The running time depends on the distribution of splits

Best Case

- If we are lucky, Partition splits the array evenly

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

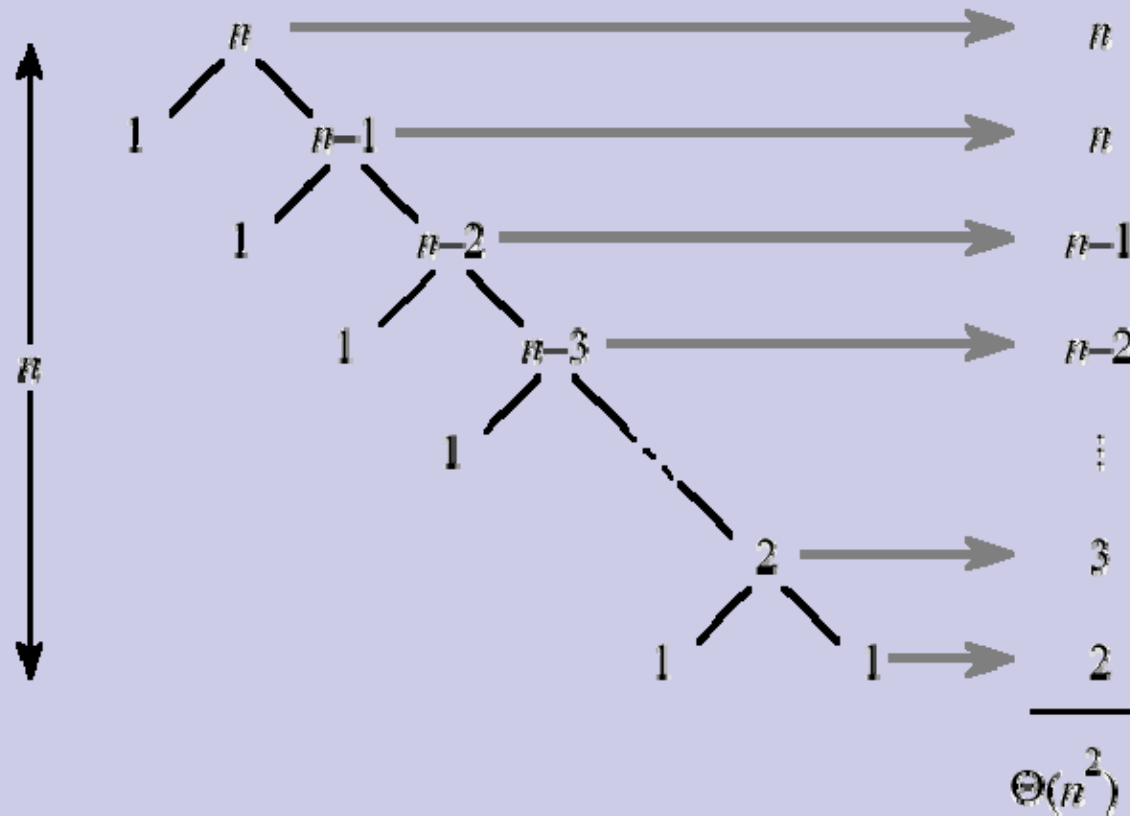


Worst Case

- What is the worst case?
- One side of the partition has only one element

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

Worst Case (2)



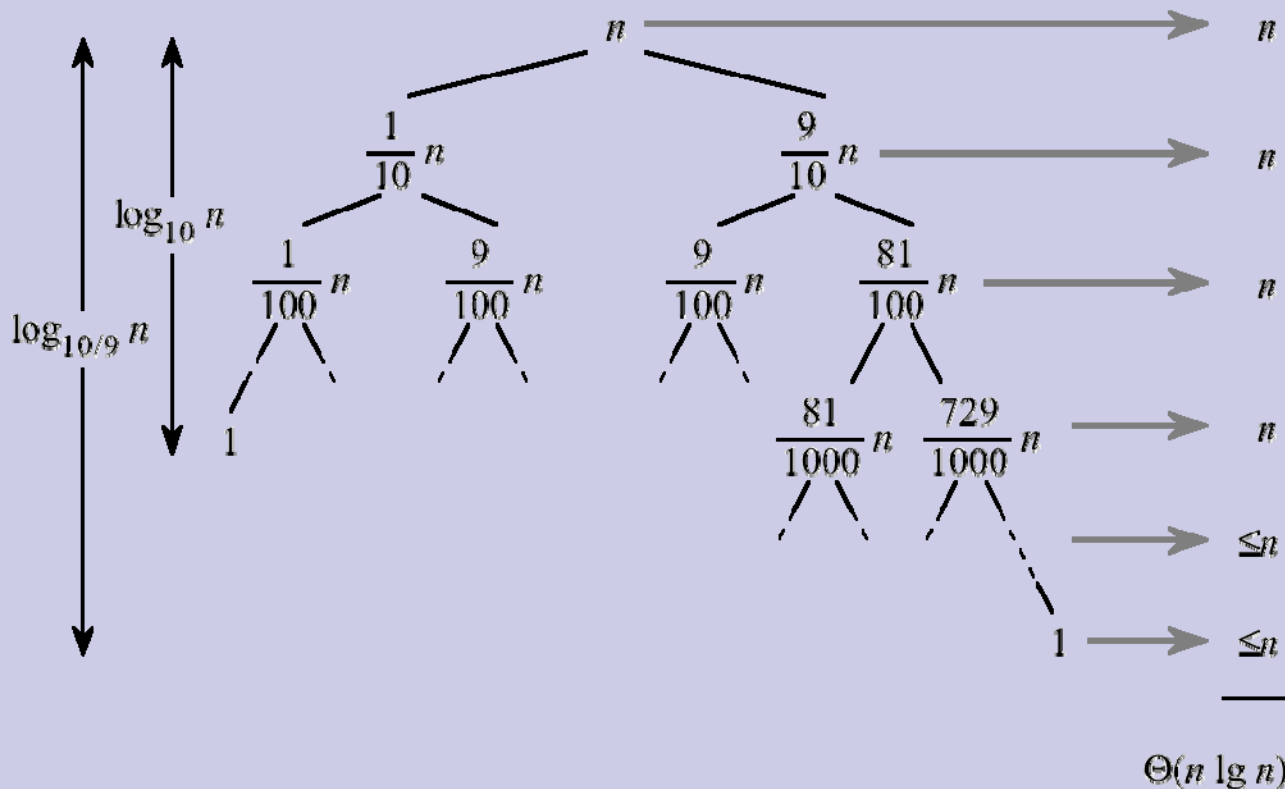
Worst Case (3)

- When does the worst case appear?
 - input is sorted
 - input reverse sorted
- Same recurrence for the worst case of insertion sort
- However, sorted input yields the best case for insertion sort!

Analysis of Quicksort

- Suppose the split is 1/10 : 9/10

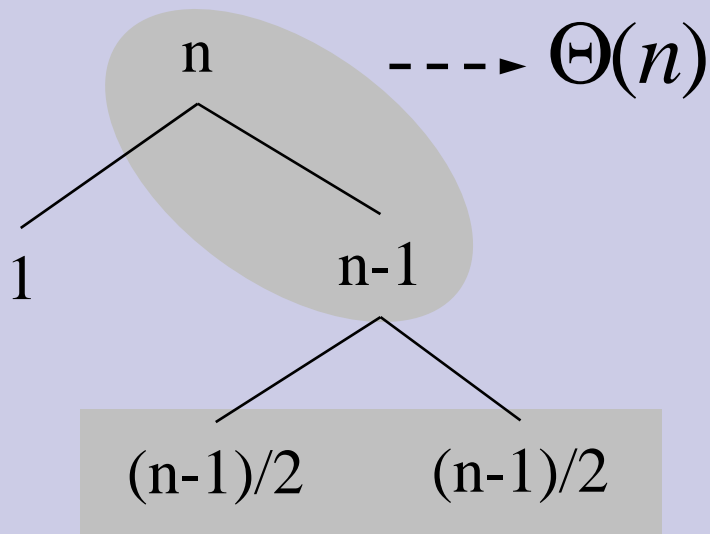
$$T(n) = T(n/10) + T(9n/10) + \Theta(n) = \Theta(n \log n)!$$





An Average Case Scenario

- Suppose, we alternate lucky and unlucky cases to get an average behavior



$$L(n) = 2U(n/2) + \Theta(n) \quad \text{lucky}$$

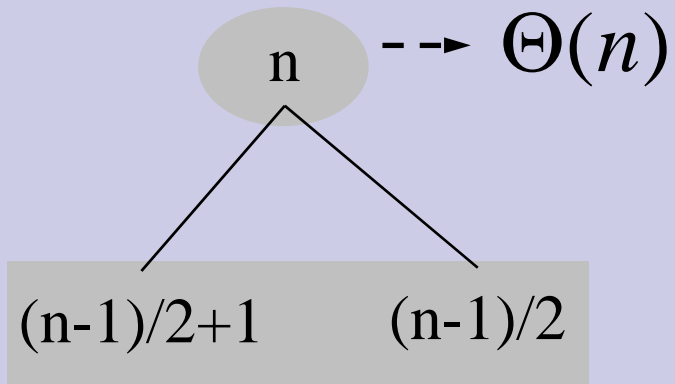
$$U(n) = L(n-1) + \Theta(n) \quad \text{unlucky}$$

we consequently get

$$L(n) = 2(L(n/2 - 1) + \Theta(n/2)) + \Theta(n)$$

$$= 2L(n/2 - 1) + \Theta(n)$$

$$= \Theta(n \log n)$$



An Average Case Scenario (2)

- How can we make sure that we are usually lucky?
 - Partition around the "middle" ($n/2$ th) element?
 - Partition around a random element (works well in practice)
- Randomized algorithm
 - running time is independent of the input ordering
 - no specific input triggers worst-case behavior
 - the worst-case is only determined by the output of the random-number generator

Randomized Quicksort

- Assume all elements are distinct
- Partition around a random element
- Consequently, all splits ($1:n-1$, $2:n-2$, ..., $n-1:1$) are equally likely with probability $1/n$
- Randomization is a general tool to improve algorithms with bad worst-case but good average-case complexity

Randomized Quicksort (2)

Randomized-Partition(A,p,r)

```
01  i ← Random(p,r)
02  exchange A[r] ↔ A[i]
03  return Partition(A,p,r)
```

Randomized-Quicksort(A,p,r)

```
01  if p < r then
02    q ← Randomized-Partition(A,p,r)
03    Randomized-Quicksort(A,p,q)
04    Randomized-Quicksort(A,q+1,r)
```

Randomized Quicksort Analysis

- Let $T(n)$ be the expected number of comparisons needed to quicksort n numbers.
- Since each split occurs with probability $1/n$, $T(n)$ has value $T(i-1)+T(n-i)+n-1$ with probability $1/n$.
- Hence,

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

Randomized Quicksort Analysis(2)

- We have seen this recurrence before.
- It is the recurrence for the expected number of comparisons required to insert a randomly chosen permutation of n elements.
- We proved that $T(n) = O(n \log_2 n)$.
- Hence expected number of comparisons required by randomized quicksort is $O(n \log_2 n)$

Randomized Quicksort running times

- Worst case running time of quicksort is $O(n^2)$
- Best case running time of quicksort is $O(n \log_2 n)$
- Expected running time of quicksort is $O(n \log_2 n)$

What does expected running time mean?

- The running time of quicksort does not depend on the input. It depends on the random numbers provided by the generator.
- Thus for the same input the program might take 3sec today and 5sec tomorrow.
- The average time taken over many different runs of the program would give us the expected time.
- Same as saying that we are taking average over all possible random number sequences provided by the generator.

Analysis of insertion in BST

- When creating a binary search tree on n elements the running time does depend on the order of the elements.
- Our algorithm for insertion did not employ an random bits.
- Given a specific input order the algorithm takes the same time each day.
- However, the time taken is different for different input orders.
- The average time taken over all possible input orders is $O(n \log_2 n)$.