Lecture 4: QuickSort

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 $\le x \le$ elements in upper subarray.



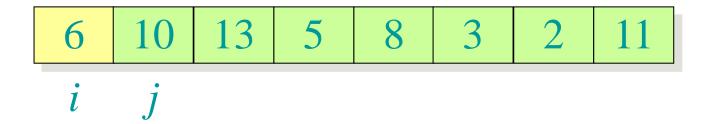
- 2. Conquer: Recursively sort the two subarrays.
- 3. Combine: Trivial.

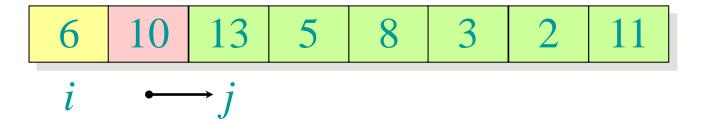
Key: Linear-time partitioning subroutine.

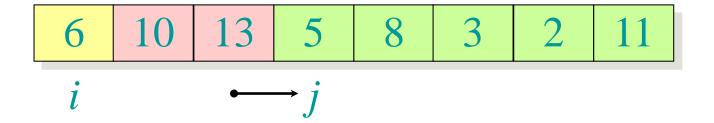
PARTITIONING SUBROUTINE

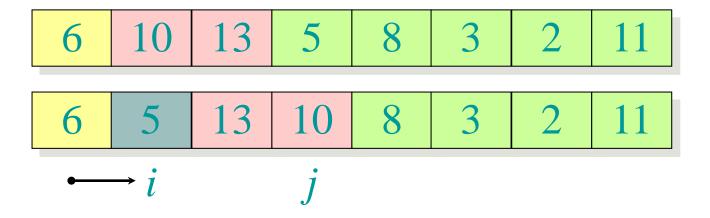
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Partition(A, p, q) \triangleright A[p ... q]
x \leftarrow A[p] \triangleright \text{pivot} = A[p]
Running time
i \leftarrow p
\text{for } j \leftarrow p+1 \text{ to } q
\text{do if } A[j] \leq x
\text{then } i \leftarrow i+1
\text{exchange } A[i] \leftrightarrow A[j]
\text{exchange } A[p] \leftrightarrow A[i]
\text{return } i
```

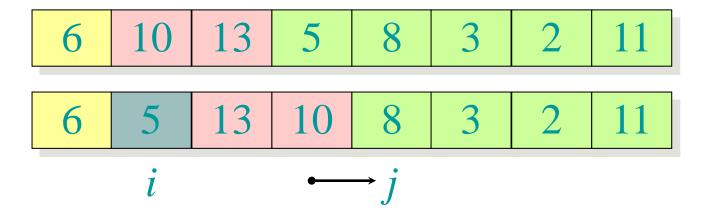
Invariant: $x \le x \ge x$? $p \qquad i \qquad j \qquad q$

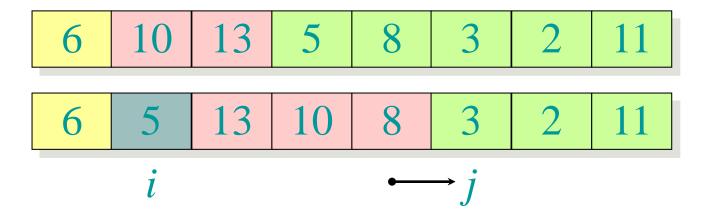


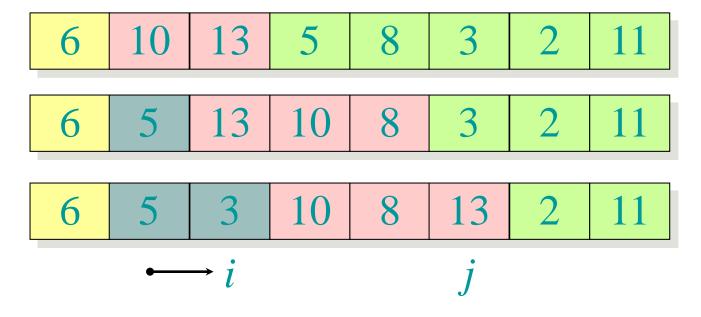


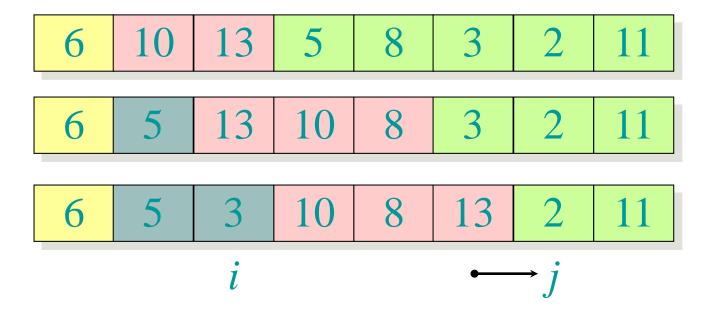












6	10	13	5	8	3	2	11
6	5	13	10	8	3	2	11
6	5	3	10	8	13	2	11
6	5	3	2	8	13	10	11
$\longrightarrow i$						\overline{j}	

6	10	13	5	8	3	2	11
6	5	13	10	8	3	2	11
6	5	3	10	8	13	2	11
6	5	3	2	8	13	10	11
i						•	$\rightarrow j$

6	10	13	5	8	3	2	11
6	5	13	10	8	3	2	11
6	5	3	10	8	13	2	11
6	5	3	2	8	13	10	11
			i				•

6	10	13	5	8	3	2	11
6	5	13	10	8	3	2	11
6	5	3	10	8	13	2	11
6	5	3	2	8	13	10	11
2	5	3	6	8	13	10	11

PSEUDOCODE FOR QUICKSORT

```
Quicksort(A, p, r)

if p < r

then q \leftarrow \text{Partition}(A, p, r)

Quicksort(A, p, q)

Quicksort(A, p, q)
```

Initial call: QUICKSORT(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.

$$T(n) = T(0) + T(n-1) + \Theta(n)$$

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

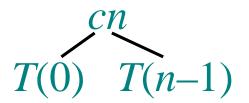
$$= T(n-1) + \Theta(n)$$

$$= \Theta(n^2) \qquad (arithmetic series)$$

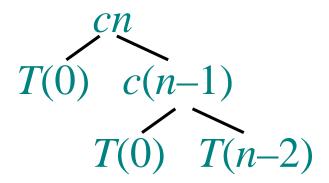
WORST-CASE RECURSION TREE

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

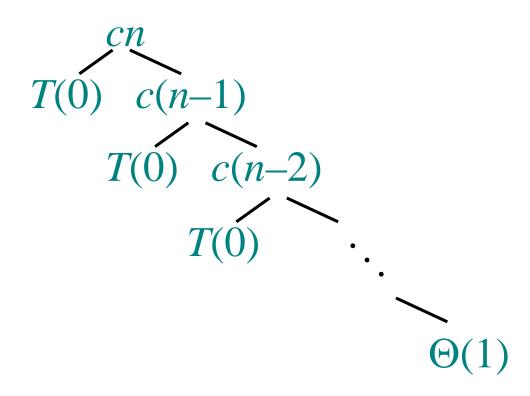
T(n)



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

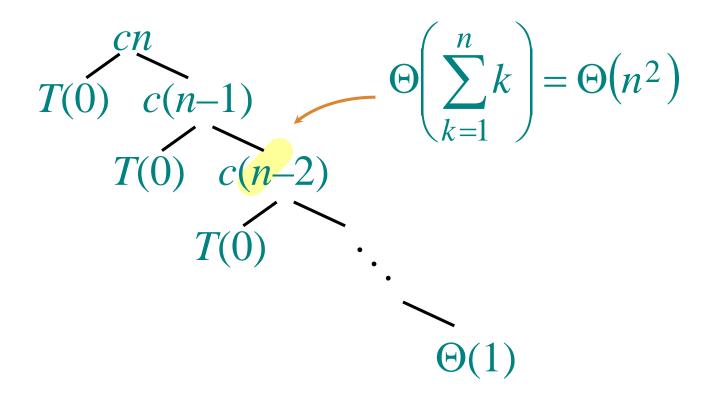


$$T(n) = T(0) + T(n-1) + cn$$

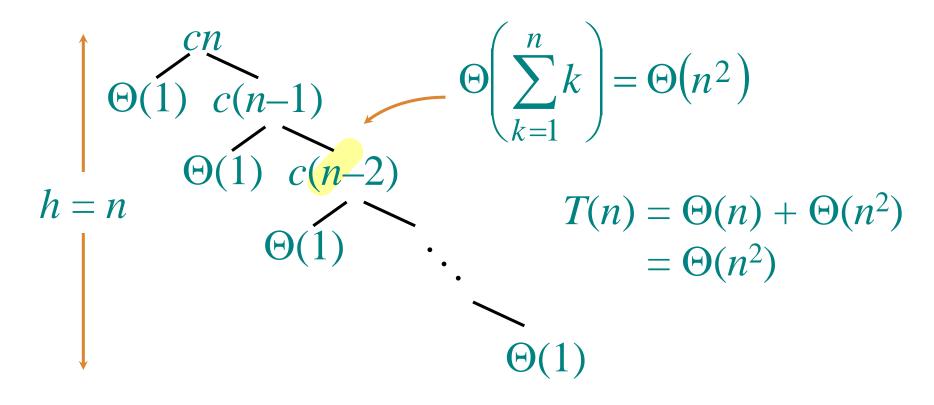


WORST-CASE RECURSION TREE

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



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



BEST-CASE ANALYSIS

If we're lucky, Partition splits the array evenly:

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

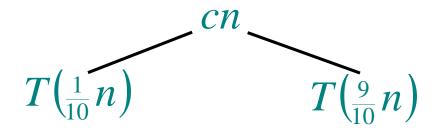
= $\Theta(n \lg n)$ (same as merge sort)

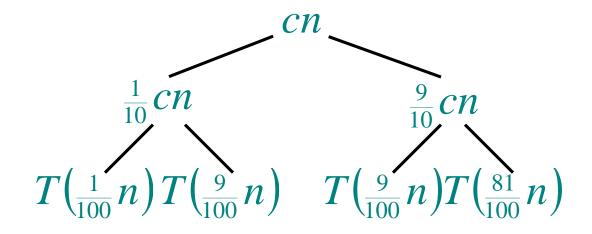
What if the split is always $\frac{1}{10}$: $\frac{9}{10}$?

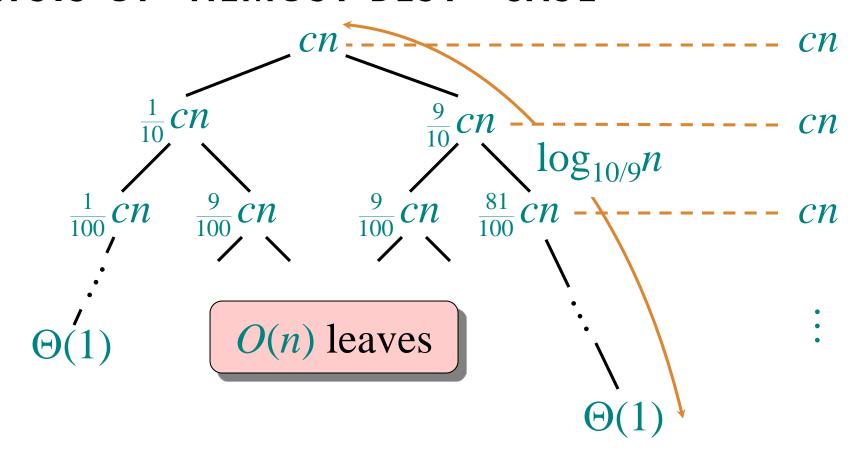
$$T(n) = T\left(\frac{1}{10}n\right) + T\left(\frac{9}{10}n\right) + \Theta(n)$$

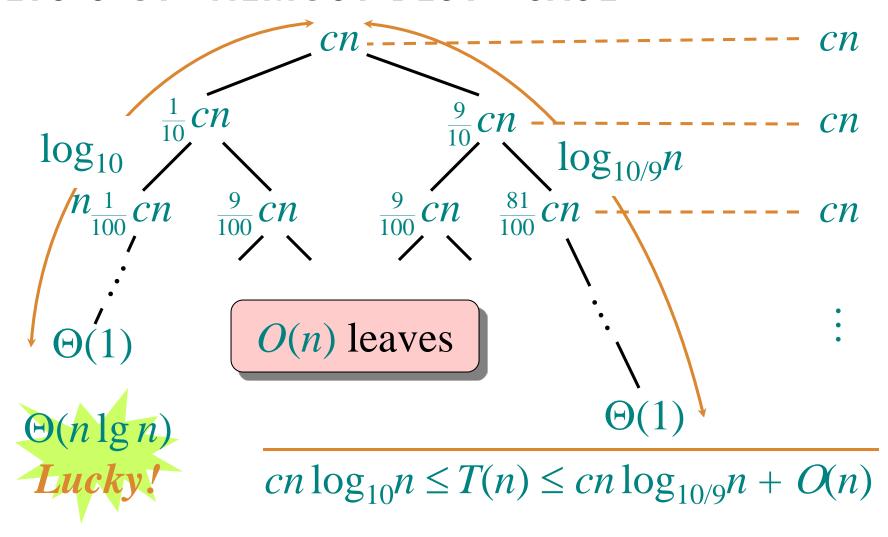
What is the solution to this recurrence?

T(n)









MORE INTUITION

Suppose we alternate lucky, unlucky, lucky, unlucky, lucky,

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

Solving:

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

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

$$= \Theta(n \lg n)$$
Lucky!

How can we make sure we are usually lucky?

RANDOMIZED QUICKSORT

IDEA: Partition around a *random* element.

- Running time is independent of the input order.
- No assumptions need to be made about the input distribution.
- No specific input elicits the worst-case behavior.
- The worst case is determined only by the output of a random-number generator.

RANDOMIZED QUICKSORT ANALYSIS

Let T(n) = the random variable for the running time of randomized quicksort on an input of size n, assuming random numbers are independent.

For k = 0, 1, ..., n-1, define the *indicator* random variable

$$X_k = \begin{cases} 1 & \text{if Partition generates a } k: n-k-1 \text{ split,} \\ 0 & \text{otherwise.} \end{cases}$$

 $E[X_k] = \Pr\{X_k = 1\} = 1/n$, since all splits are equally likely, assuming elements are distinct.

ANALYSIS (CONTINUED)

$$T(n) = \begin{cases} T(0) + T(n-1) + \Theta(n) & \text{if } 0 : n-1 \text{ split,} \\ T(1) + T(n-2) + \Theta(n) & \text{if } 1 : n-2 \text{ split,} \\ \vdots & & \vdots \\ T(n-1) + T(0) + \Theta(n) & \text{if } n-1 : 0 \text{ split,} \end{cases}$$

$$= \sum_{k=0}^{n-1} X_k \left(T(k) + T(n-k-1) + \Theta(n) \right).$$

CALCULATING EXPECTATION
$$E[T(n)] = E \left[\sum_{k=0}^{n-1} X_k (T(k) + T(n-k-1) + \Theta(n)) \right]$$

Take expectations of both sides.

CALCULATING EXPECTATION
$$E[T(n)] = E \left[\sum_{k=0}^{n-1} X_k (T(k) + T(n-k-1) + \Theta(n)) \right]$$

$$= \sum_{k=0}^{n-1} E[X_k (T(k) + T(n-k-1) + \Theta(n))]$$

Linearity of expectation.

CALCULATING EXPECTATION
$$E[T(n)] = E\left[\sum_{k=0}^{n-1} X_k (T(k) + T(n-k-1) + \Theta(n))\right]$$

$$= \sum_{k=0}^{n-1} E[X_k (T(k) + T(n-k-1) + \Theta(n))]$$

$$= \sum_{k=0}^{n-1} E[X_k] \cdot E[T(k) + T(n-k-1) + \Theta(n)]$$

Independence of X_k from other random choices.

CALCULATING EXPECTATION

$$\begin{split} E[T(n)] &= E \Bigg[\sum_{k=0}^{n-1} X_k \big(T(k) + T(n-k-1) + \Theta(n) \big) \Bigg] \\ &= \sum_{k=0}^{n-1} E \big[X_k \big(T(k) + T(n-k-1) + \Theta(n) \big) \big] \\ &= \sum_{k=0}^{n-1} E \big[X_k \big] \cdot E \big[T(k) + T(n-k-1) + \Theta(n) \big] \\ &= \frac{1}{n} \sum_{k=0}^{n-1} E \big[T(k) \big] + \frac{1}{n} \sum_{k=0}^{n-1} E \big[T(n-k-1) \big] + \frac{1}{n} \sum_{k=0}^{n-1} \Theta(n) \end{split}$$

Linearity of expectation; $E[X_k] = 1/n$.

CALCULATING EXPECTATION

$$E[T(n)] = E\left[\sum_{k=0}^{n-1} X_k (T(k) + T(n-k-1) + \Theta(n))\right]$$

$$= \sum_{k=0}^{n-1} E[X_k (T(k) + T(n-k-1) + \Theta(n))]$$

$$= \sum_{k=0}^{n-1} E[X_k] \cdot E[T(k) + T(n-k-1) + \Theta(n)]$$

$$= \frac{1}{n} \sum_{k=0}^{n-1} E[T(k)] + \frac{1}{n} \sum_{k=0}^{n-1} E[T(n-k-1)] + \frac{1}{n} \sum_{k=0}^{n-1} \Theta(n)$$

$$= \frac{2}{n} \sum_{k=1}^{n-1} E[T(k)] + \Theta(n)$$
Summations have identical terms.

HAIRY RECURRENCE

$$E[T(n)] = \frac{2}{n} \sum_{k=2}^{n-1} E[T(k)] + \Theta(n)$$

(The k = 0, 1 terms can be absorbed in the $\Theta(n)$.)

Prove: $E[T(n)] \le an \lg n$ for constant a > 0.

• Choose *a* large enough so that $an \lg n$ dominates E[T(n)] for sufficiently small $n \ge 2$.

Use fact:
$$\sum_{k=2}^{n-1} k \lg k \le \frac{1}{2} n^2 \lg n - \frac{1}{8} n^2$$
 (exercise).

$$E[T(n)] \le \frac{2}{n} \sum_{k=2}^{n-1} ak \lg k + \Theta(n)$$

Substitute inductive hypothesis.

$$E[T(n)] \le \frac{2}{n} \sum_{k=2}^{n-1} ak \lg k + \Theta(n)$$

$$\le \frac{2a}{n} \left(\frac{1}{2}n^2 \lg n - \frac{1}{8}n^2\right) + \Theta(n)$$

Use fact.

$$E[T(n)] \le \frac{2}{n} \sum_{k=2}^{n-1} ak \lg k + \Theta(n)$$

$$\le \frac{2a}{n} \left(\frac{1}{2} n^2 \lg n - \frac{1}{8} n^2 \right) + \Theta(n)$$

$$= an \lg n - \left(\frac{an}{4} - \Theta(n) \right)$$

Express as *desired* – *residual*.

$$E[T(n)] \le \frac{2}{n} \sum_{k=2}^{n-1} ak \lg k + \Theta(n)$$

$$= \frac{2a}{n} \left(\frac{1}{2} n^2 \lg n - \frac{1}{8} n^2 \right) + \Theta(n)$$

$$= an \lg n - \left(\frac{an}{4} - \Theta(n) \right)$$

$$\le an \lg n,$$

if a is chosen large enough so that an/4 dominates the $\Theta(n)$.

QUICKSORT IN PRACTICE

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