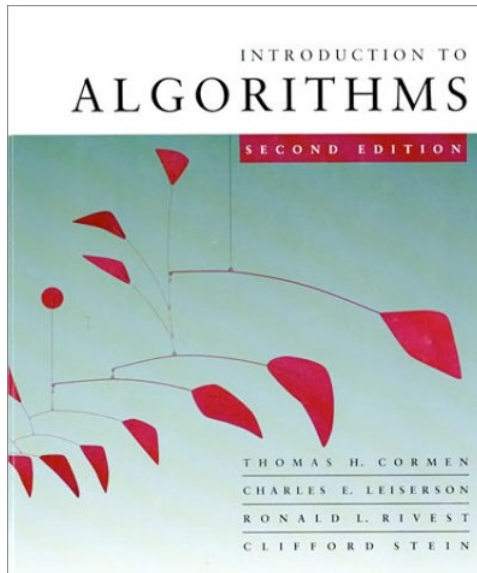


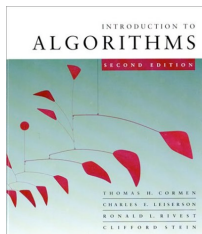
Introduction to Algorithms

6.046J/18.401J



Lecture 6

Prof. Piotr Indyk



Order statistics

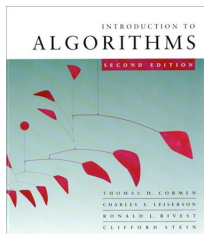
Select the i th smallest of n elements (the element with *rank* i).

- $i = 1$: *minimum*;
- $i = n$: *maximum*;
- $i = \lfloor (n+1)/2 \rfloor$ or $\lceil (n+1)/2 \rceil$: *median*.

Naive algorithm: Sort and index i th element.

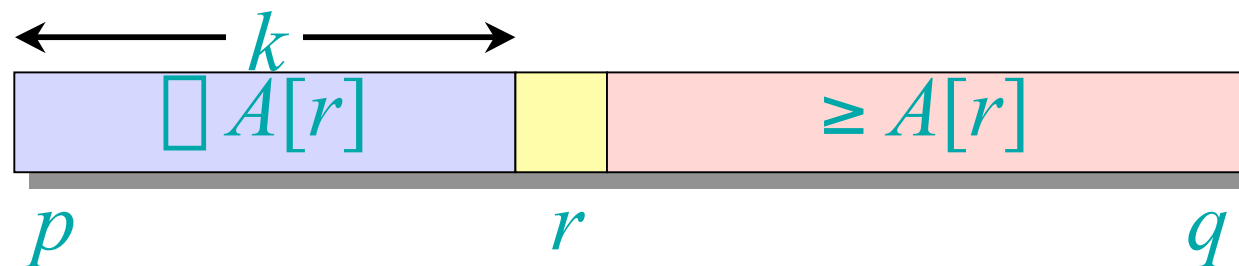
$$\begin{aligned}\text{Worst-case running time} &= \Theta(n \lg n) + \Theta(1) \\ &= \Theta(n \lg n),\end{aligned}$$

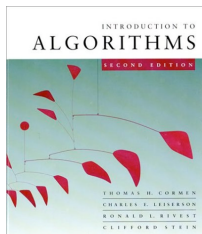
using merge sort.



Randomized divide-and-conquer

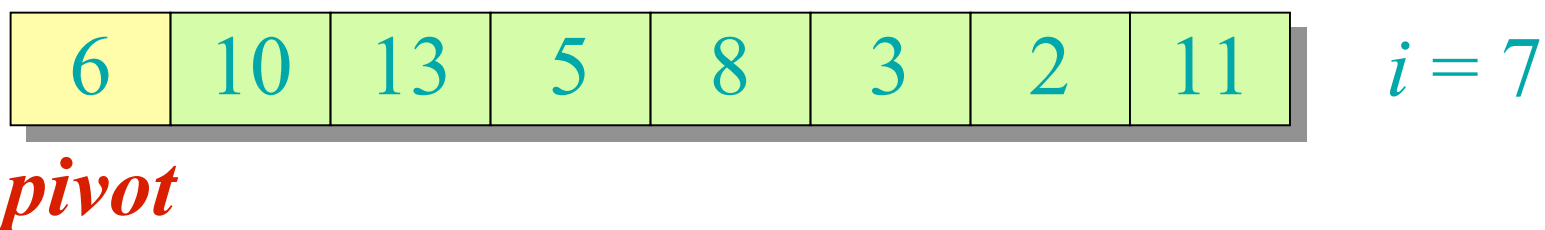
RAND-SELECT(A, p, q, i) i th smallest of $A[p..q]$
if $p = q$ **then return** $A[p]$
 $r \leftarrow$ **RAND-PARTITION**(A, p, q)
 $k \leftarrow r - p + 1$ $k = \text{rank}(A[r])$
if $i = k$ **then return** $A[r]$
if $i < k$
 then return **RAND-SELECT**($A, p, r - 1, i$)
 else return **RAND-SELECT**($A, r + 1, q, i - k$)



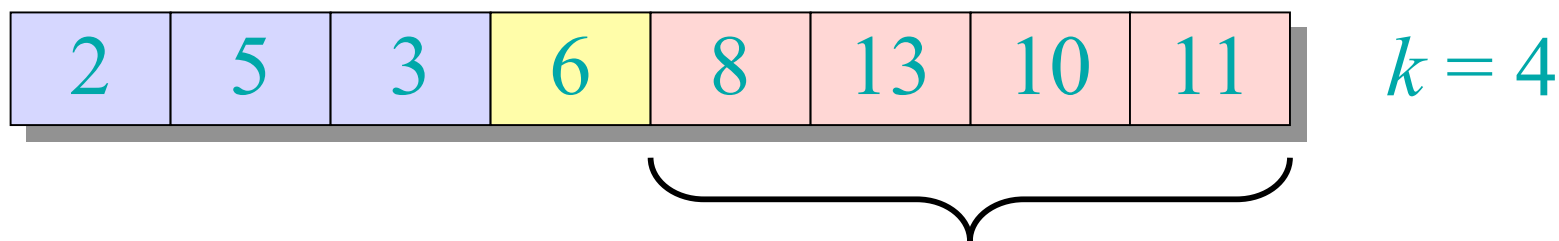


Example

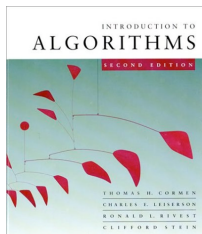
Select the $i = 7$ th smallest:



Partition:



Select the $7 - 4 = 3$ rd smallest recursively.



Intuition for analysis

(All our analyses today assume that all elements are distinct.)

Lucky:

$$\begin{aligned} T(n) &= T(9n/10) + \Theta(n) \\ &= \Theta(n) \end{aligned}$$

$$n^{\log_{10/9} 1} = n^0 = 1$$

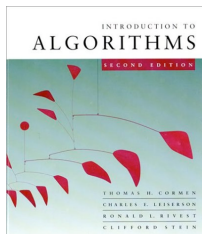
CASE 3

Unlucky:

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

arithmetic series

Worse than sorting!



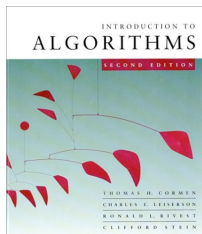
Analysis

- The probability that a random pivot induces lucky partition is at least $8/10$ (Lecture 4)
- Let t_i be the number of partitions performed between the $(i-1)$ -th and the i -th lucky partition
- The total time is at most:

$$t_1 n + t_2 (9/10) n + t_3 (9/10)^2 n + \dots$$

- The total **expected** time is at most:

$$10/8 n + 10/8 (9/10) n + 10/8 (9/10)^2 n + \dots \\ = O(n)$$



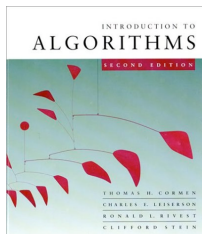
Alternative analysis of expected time

The analysis follows that of randomized quicksort, but it's a little different.

Let $T(n)$ = the random variable for the running time of RAND-SELECT on an input of size n , assuming random numbers are independent.

For $k = 0, 1, \dots, n-1$, define the *indicator random variable*

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

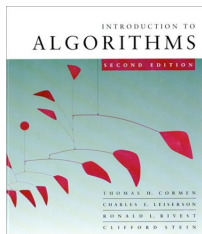


Analysis (continued)

To obtain an upper bound, assume that the i th element always falls in the larger side of the partition:

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

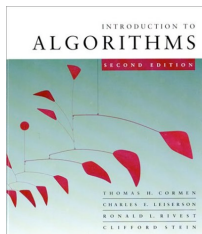
$$= \sum_{k=0}^{n-1} X_k (T(\max\{k, n-k-1\}) + \Theta(n)).$$



Calculating expectation

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

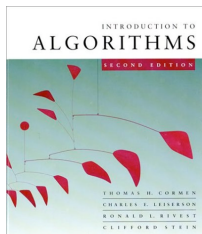
Take expectations of both sides.



Calculating expectation

$$\begin{aligned} E[T(n)] &= E \left[\sum_{k=0}^{n-1} X_k (T(\max\{k, n-k-1\}) + \Theta(n)) \right] \\ &= \sum_{k=0}^{n-1} E[X_k (T(\max\{k, n-k-1\}) + \Theta(n))] \end{aligned}$$

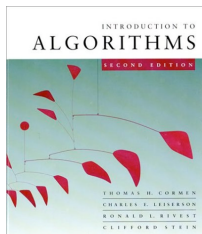
Linearity of expectation.



Calculating expectation

$$\begin{aligned}
 E[T(n)] &= E\left[\sum_{k=0}^{n-1} X_k (T(\max\{k, n-k-1\}) + \Theta(n))\right] \\
 &= \sum_{k=0}^{n-1} E[X_k (T(\max\{k, n-k-1\}) + \Theta(n))] \\
 &= \sum_{k=0}^{n-1} E[X_k] \cdot E[T(\max\{k, n-k-1\}) + \Theta(n)]
 \end{aligned}$$

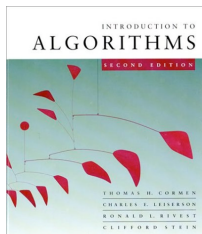
Independence of X_k from other random choices.



Calculating expectation

$$\begin{aligned} E[T(n)] &= E\left[\sum_{k=0}^{n-1} X_k (T(\max\{k, n-k-1\}) + \Theta(n))\right] \\ &= \sum_{k=0}^{n-1} E[X_k (T(\max\{k, n-k-1\}) + \Theta(n))] \\ &= \sum_{k=0}^{n-1} E[X_k] \cdot E[T(\max\{k, n-k-1\}) + \Theta(n)] \\ &= \frac{1}{n} \sum_{k=0}^{n-1} E[T(\max\{k, n-k-1\})] + \frac{1}{n} \sum_{k=0}^{n-1} \Theta(n) \end{aligned}$$

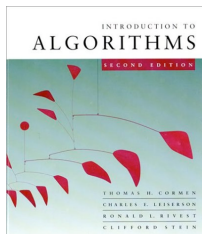
Linearity of expectation; $E[X_k] = 1/n$.



Calculating expectation

$$\begin{aligned}
 E[T(n)] &= E \left[\sum_{k=0}^{n-1} X_k (T(\max\{k, n-k-1\}) + \Theta(n)) \right] \\
 &= \sum_{k=0}^{n-1} E[X_k (T(\max\{k, n-k-1\}) + \Theta(n))] \\
 &= \sum_{k=0}^{n-1} E[X_k] \cdot E[T(\max\{k, n-k-1\}) + \Theta(n)] \\
 &= \frac{1}{n} \sum_{k=0}^{n-1} E[T(\max\{k, n-k-1\})] + \frac{1}{n} \sum_{k=0}^{n-1} \Theta(n) \\
 &= \frac{2}{n} \sum_{k=\lceil n/2 \rceil}^{n-1} E[T(k)] + \Theta(n)
 \end{aligned}$$

Upper terms
appear twice.



Hairy recurrence

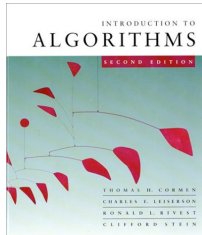
(But not quite as hairy as the quicksort one.)

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

Prove: $E[T(n)] \leq cn$ for constant $c > 0$.

- The constant c can be chosen large enough so that $E[T(n)] \leq cn$ for the base cases.

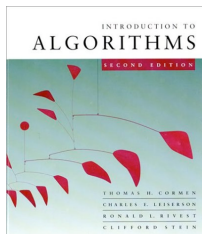
Use fact: $\sum_{k=\lfloor n/2 \rfloor}^{n-1} k \leq \frac{3}{8}n^2$ (exercise).



Substitution method

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

Substitute inductive hypothesis.

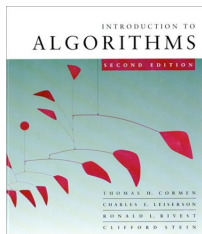


Substitution method

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

$$= \frac{2c}{n} \left[\frac{3}{8} n^2 \right] + \Theta(n)$$

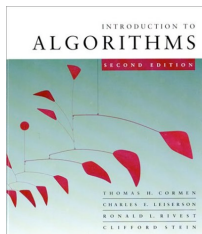
Use fact.



Substitution method

$$\begin{aligned}
 E[T(n)] &= \frac{2}{n} \sum_{k=\lfloor n/2 \rfloor}^{n-1} ck + \Theta(n) \\
 &= \frac{2c}{n} \left[\frac{3}{8} n^2 \right] + \Theta(n) \\
 &= cn \left[\frac{cn}{4} \right] \Theta(n)
 \end{aligned}$$

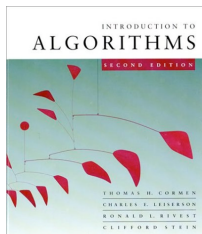
Express as *desired – residual*.



Substitution method

$$\begin{aligned}
 E[T(n)] &= \sum_{k=\lfloor n/2 \rfloor}^{n-1} 2ck + \Theta(n) \\
 &= \frac{2c}{n} \sum_{k=\lfloor n/2 \rfloor}^{n-1} k + \Theta(n) \\
 &= \frac{2c}{n} \left(\frac{(n-1) + \lfloor n/2 \rfloor}{2} \cdot (n - \lfloor n/2 \rfloor) \right) + \Theta(n) \\
 &= \frac{c}{4} n + \Theta(n) \\
 &= \Theta(n),
 \end{aligned}$$

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



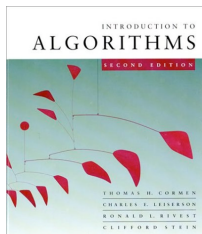
Summary of randomized order-statistic selection

- Works fast: linear expected time.
- Excellent algorithm in practice.
- But, the worst case is *very* bad: $\Omega(n^2)$.

Q. Is there an algorithm that runs in linear time in the worst case?

A. Yes, due to Blum, Floyd, Pratt, Rivest, and Tarjan [1973].

IDEA: Generate a good pivot recursively.

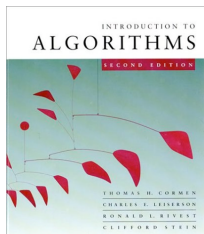


Worst-case linear-time order statistics

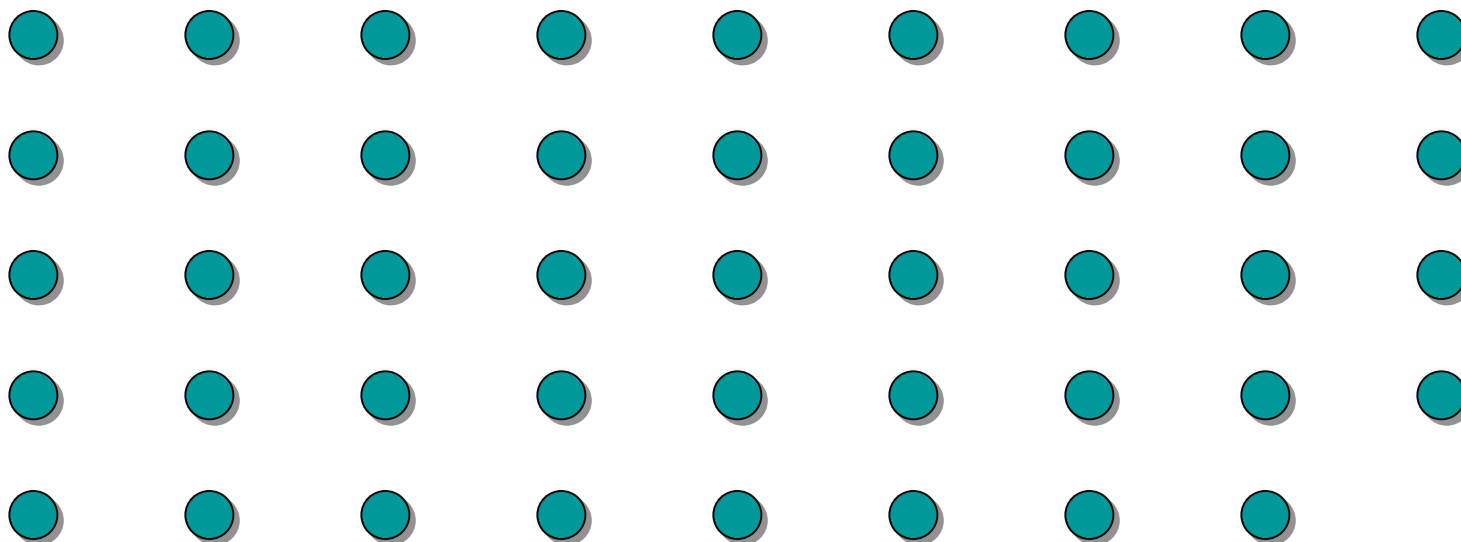
SELECT(i, n)

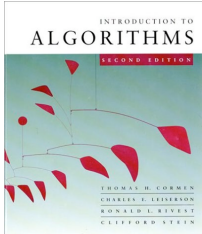
1. Divide the n elements into groups of 5. Find the median of each 5-element group by rote.
2. Recursively SELECT the median x of the $\lfloor n/5 \rfloor$ group medians to be the pivot.
3. Partition around the pivot x . Let $k = \text{rank}(x)$.
4. **if** $i = k$ **then return** x
elseif $i < k$
then recursively SELECT the i th smallest element in the lower part
else recursively SELECT the $(i-k)$ th smallest element in the upper part

Same as
RAND-
SELECT

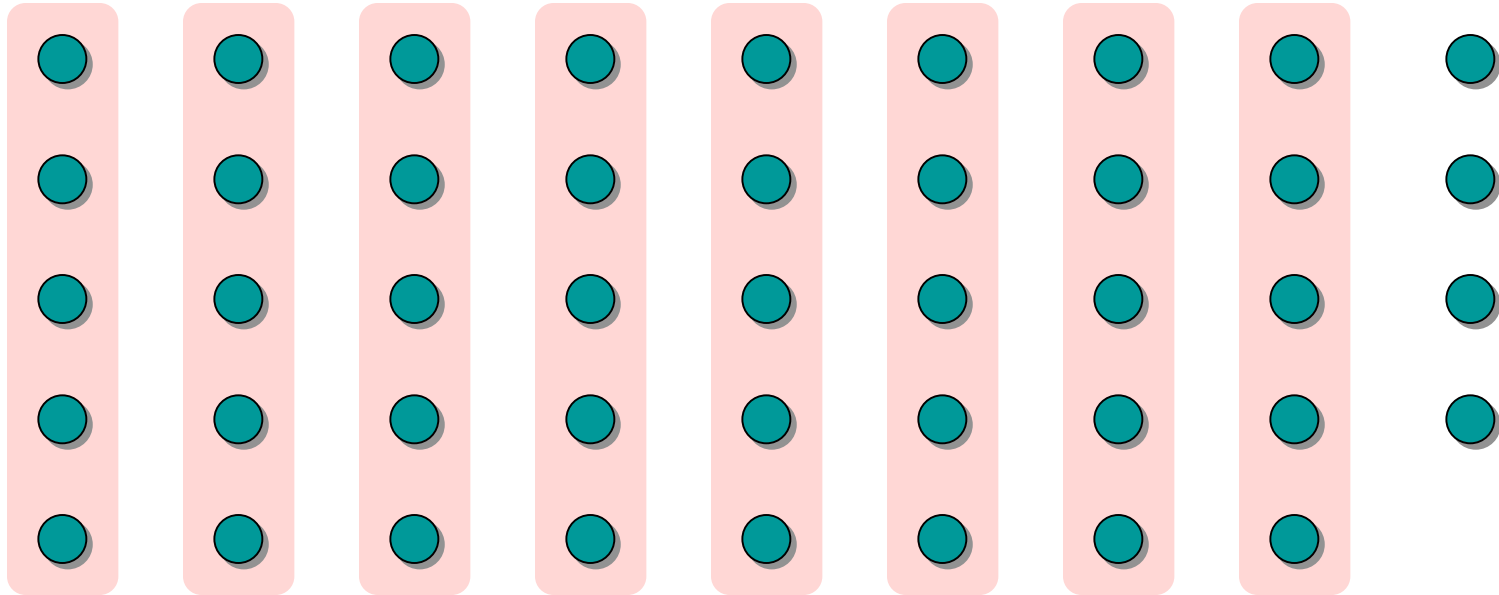


Choosing the pivot

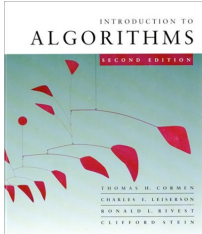




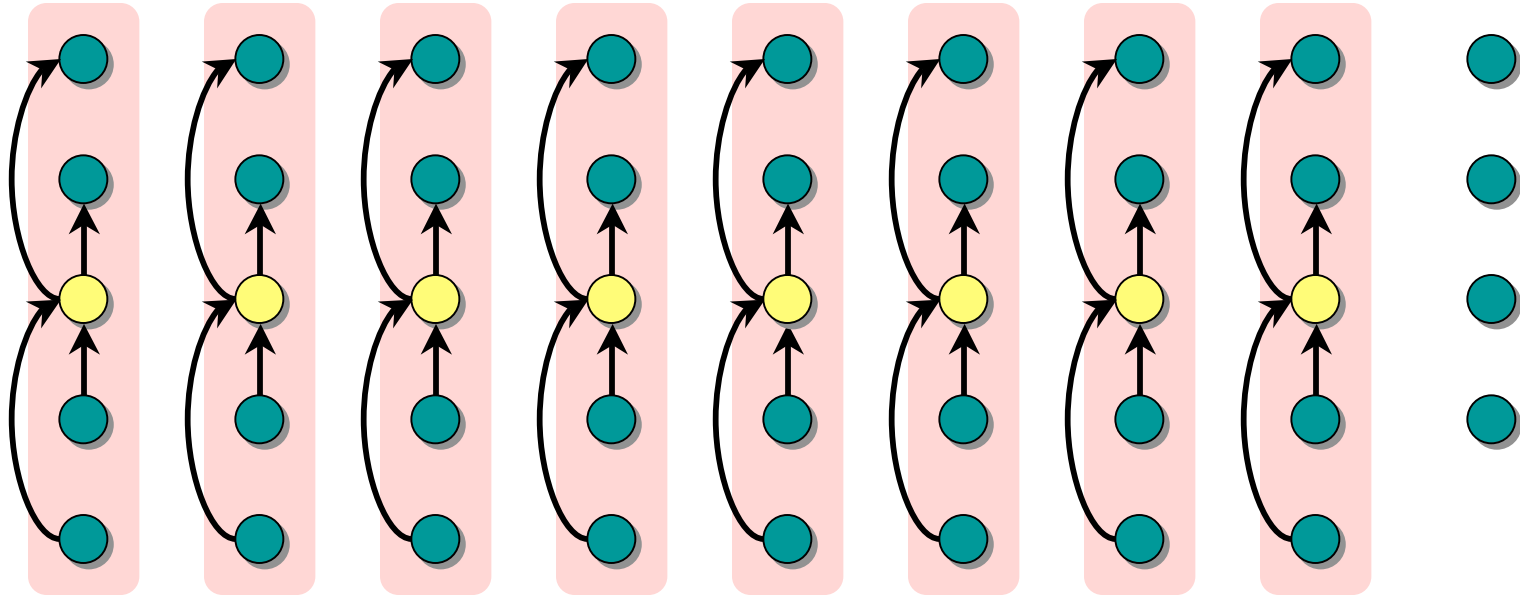
Choosing the pivot



1. Divide the n elements into groups of 5.



Choosing the pivot

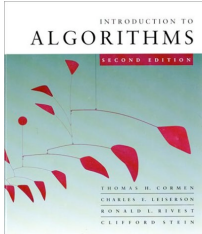


1. Divide the n elements into groups of 5. Find the median of each 5-element group by rote.

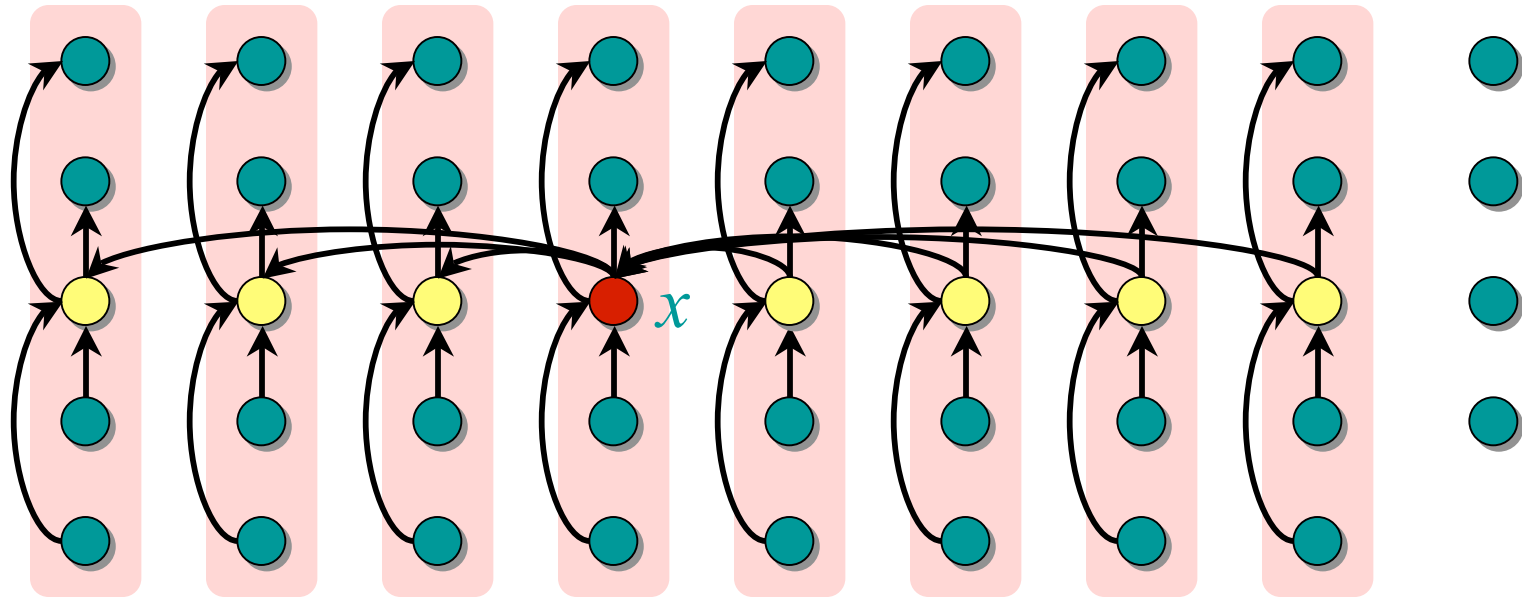
lesser



greater



Choosing the pivot

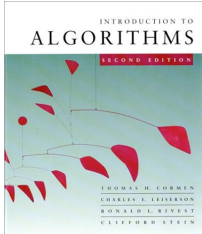


1. Divide the n elements into groups of 5. Find the median of each 5-element group by rote.
2. Recursively SELECT the median x of the $\lfloor n/5 \rfloor$ group medians to be the pivot.

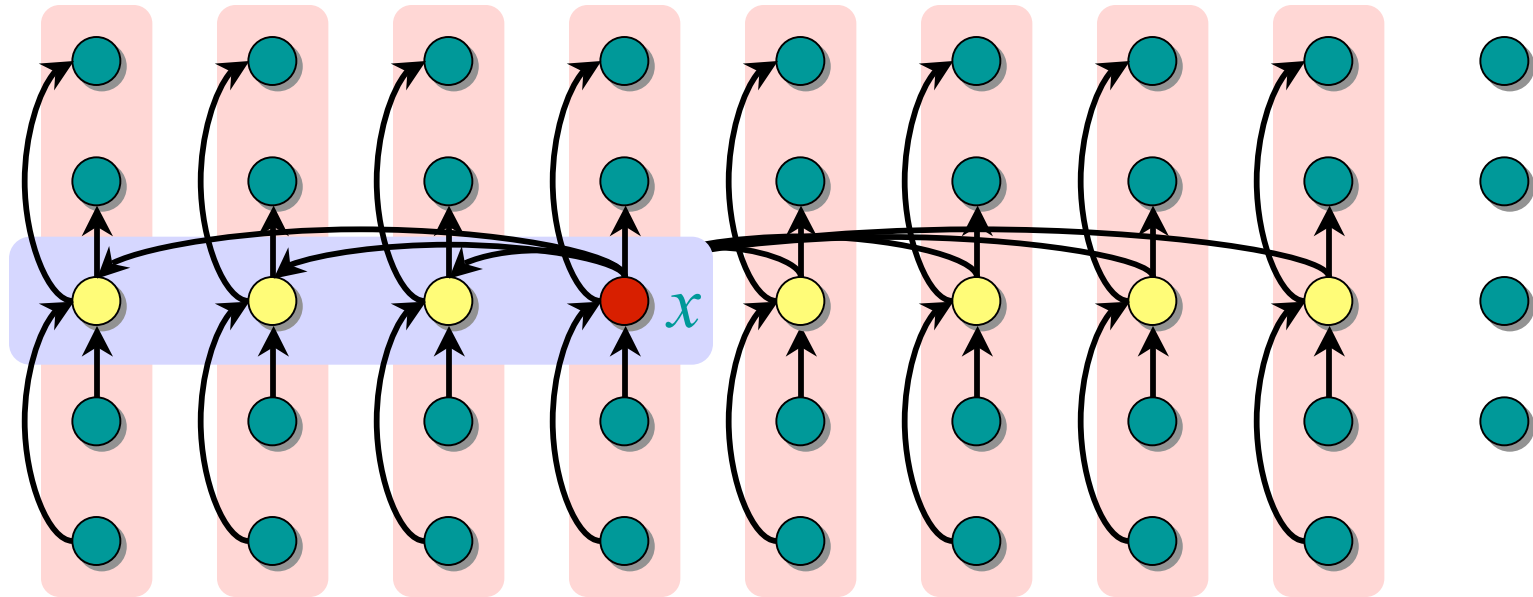
lesser



greater



Analysis

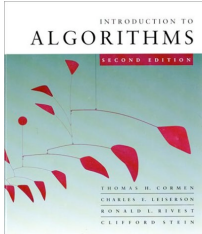


At least half the group medians are $\leq x$, which is at least $\lfloor \lfloor n/5 \rfloor / 2 \rfloor = \lfloor n/10 \rfloor$ group medians.

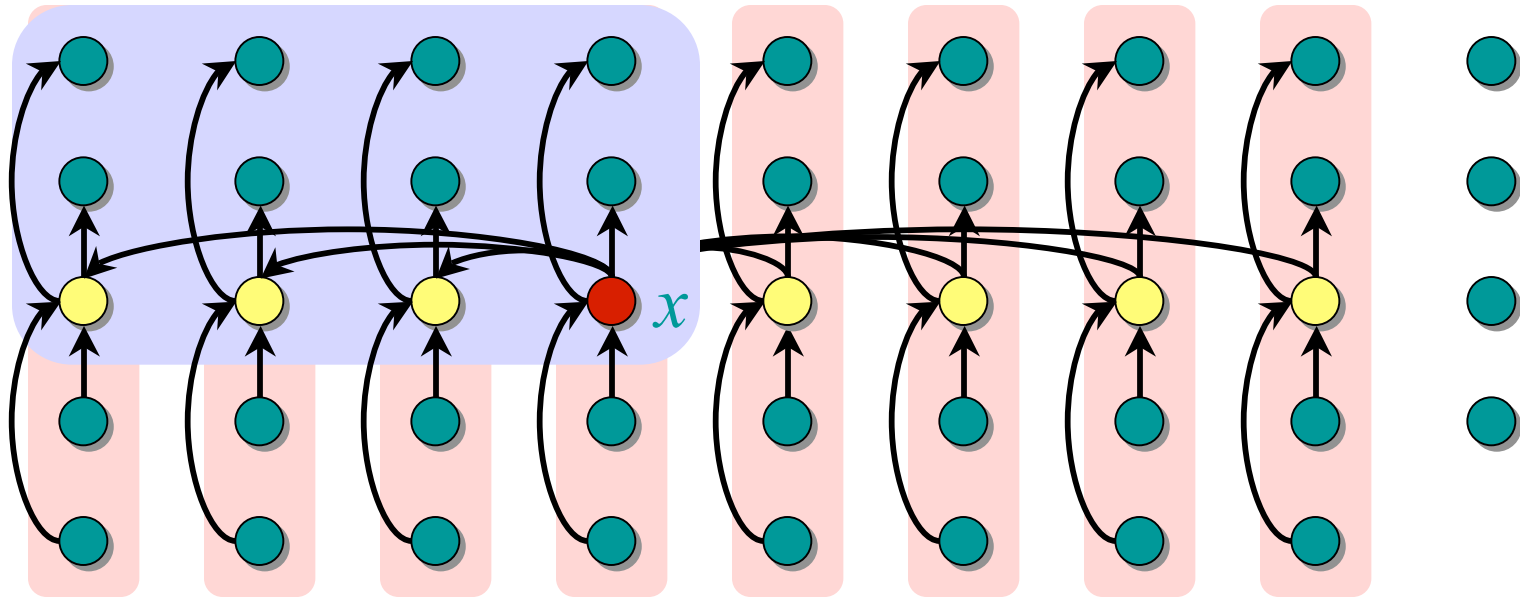
lesser



greater



Analysis (Assume all elements are distinct.)



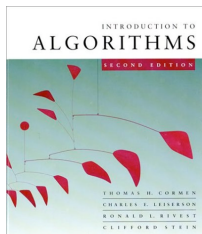
At least half the group medians are $\leq x$, which is at least $\lfloor \lfloor n/5 \rfloor / 2 \rfloor = \lfloor n/10 \rfloor$ group medians.

- Therefore, at least $3 \lfloor n/10 \rfloor$ elements are $\leq x$.

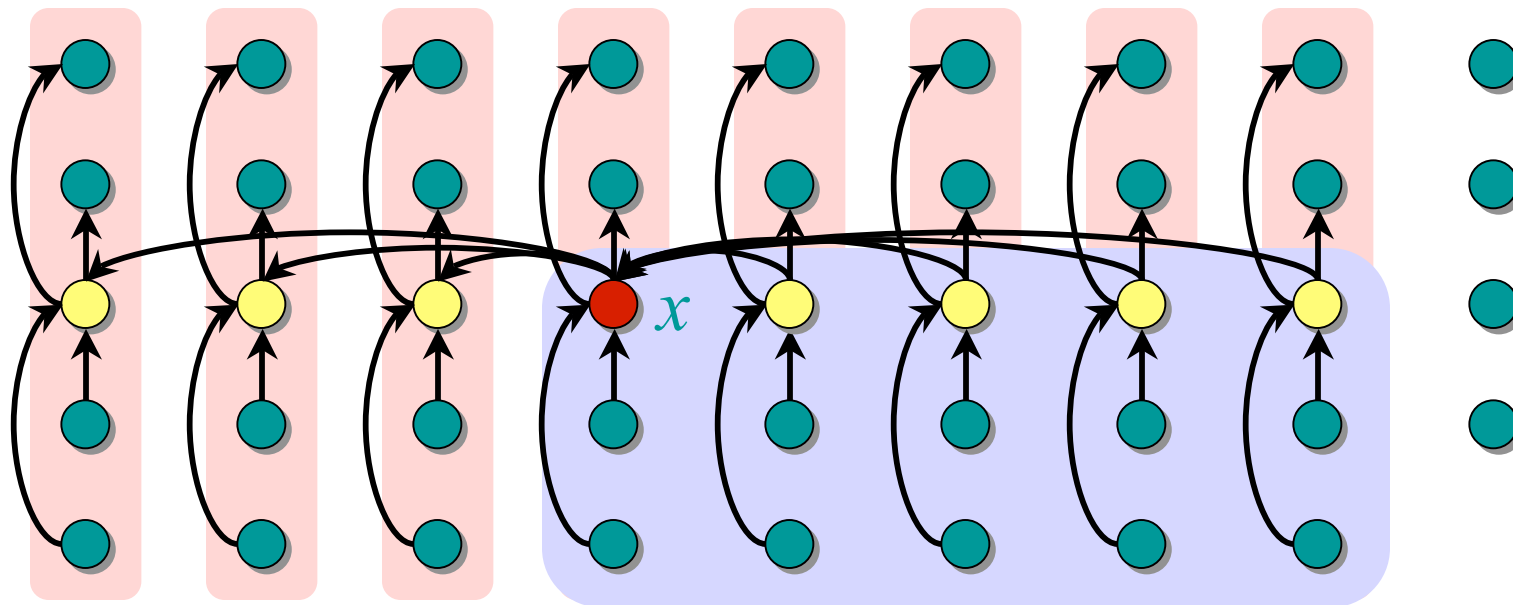
lesser



greater



Analysis (Assume all elements are distinct.)



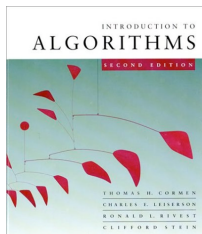
At least half the group medians are $\leq x$, which is at least $\lfloor \lfloor n/5 \rfloor / 2 \rfloor = \lfloor n/10 \rfloor$ group medians.

- Therefore, at least $3 \lfloor n/10 \rfloor$ elements are $\leq x$.
- Similarly, at least $3 \lfloor n/10 \rfloor$ elements are $\geq x$.

lesser

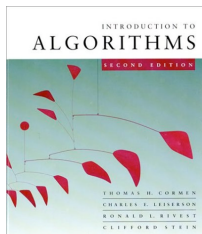


greater



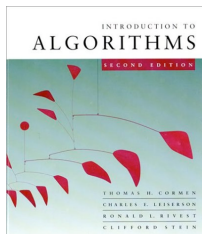
Minor simplification

- For $n \geq 50$, we have $3 \lfloor n/10 \rfloor \geq n/4$.
- Therefore, for $n \geq 50$ the recursive call to SELECT in Step 4 is executed recursively on $\lfloor 3n/4 \rfloor$ elements.
- Thus, the recurrence for running time can assume that Step 4 takes time $T(3n/4)$ in the worst case.
- For $n < 50$, we know that the worst-case time is $T(n) = \Theta(1)$.



Developing the recurrence

$T(n)$	SELECT(i, n)
$\square(n)$	1. Divide the n elements into groups of 5. Find the median of each 5-element group by rote.
$T(n/5)$	
$\square(n)$	3. Partition around the pivot x . Let $k = \text{rank}(x)$.
$T(3n/4)$	4. if $i = k$ then return x elseif $i < k$ then recursively SELECT the i th smallest element in the lower part else recursively SELECT the $(i-k)$ th smallest element in the upper part



Solving the recurrence

$$T(n) = T\left(\frac{1}{5}n\right) + T\left(\frac{3}{4}n\right) + \Theta(n)$$

Substitution:

$$T(n) \leq cn$$

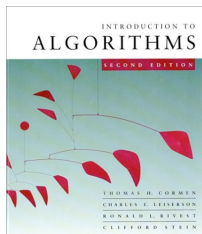
$$T(n) \leq \frac{1}{5}cn + \frac{3}{4}cn + \Theta(n)$$

$$= \frac{19}{20}cn + \Theta(n)$$

$$= cn - \frac{1}{20}cn + \Theta(n)$$

$$\leq cn,$$

if c is chosen large enough to handle both the $\Theta(n)$ and the initial conditions.



Conclusions

- Since the work at each level of recursion is a constant fraction ($19/20$) smaller, the work per level is a geometric series dominated by the linear work at the root.
- In practice, this algorithm runs slowly, because the constant in front of n is large.
- The randomized algorithm is far more practical.

Exercise: *Why not divide into groups of 3?*