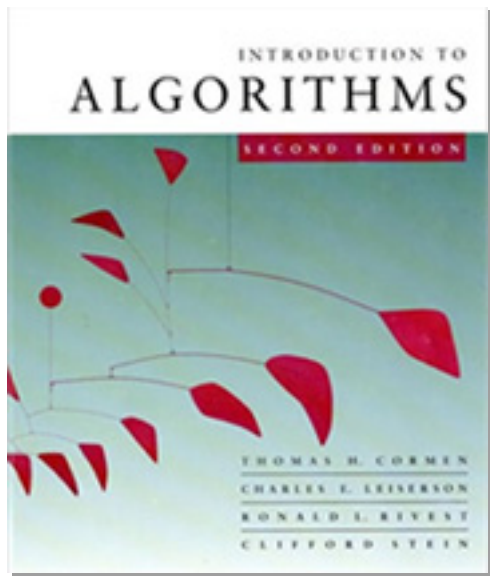


# *Introduction to Algorithms*

6.046J/18.401J

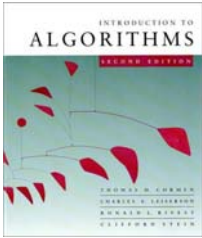


## LECTURE 6

### Order Statistics

- Randomized divide and conquer
- Analysis of expected time
- Worst-case linear-time order statistics
- Analysis

**Prof. Erik Demaine**



# Order statistics

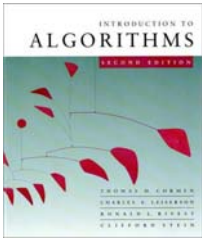
Select the  $i$ th smallest of  $n$  elements (the element with *rank*  $i$ ).

- $i = 1$ : *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.

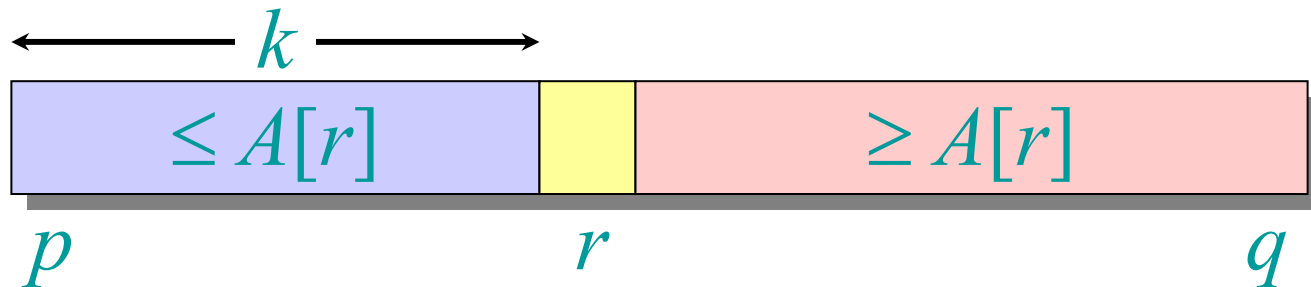
Worst-case running time =  $\Theta(n \lg n) + \Theta(1)$   
=  $\Theta(n \lg n)$ ,

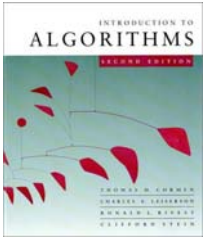
using merge sort or heapsort (*not* quicksort).



# Randomized divide-and-conquer algorithm

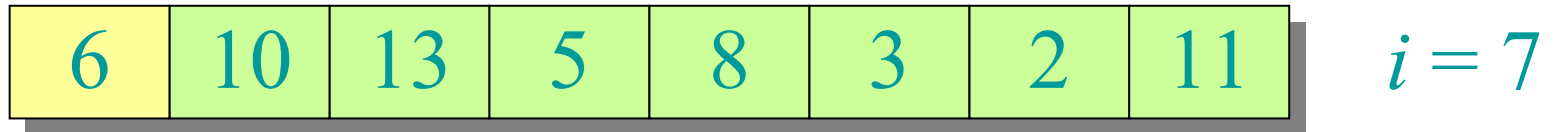
**RAND-SELECT**( $A, p, q, i$ )  $\triangleright$   $i$ th smallest of  $A[p..q]$   
**if**  $p = q$  **then return**  $A[p]$   
 $r \leftarrow$  **RAND-PARTITION**( $A, p, q$ )  
 $k \leftarrow r - p + 1$   $\triangleright k = \text{rank}(A[r])$   
**if**  $i = k$  **then return**  $A[r]$   
**if**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





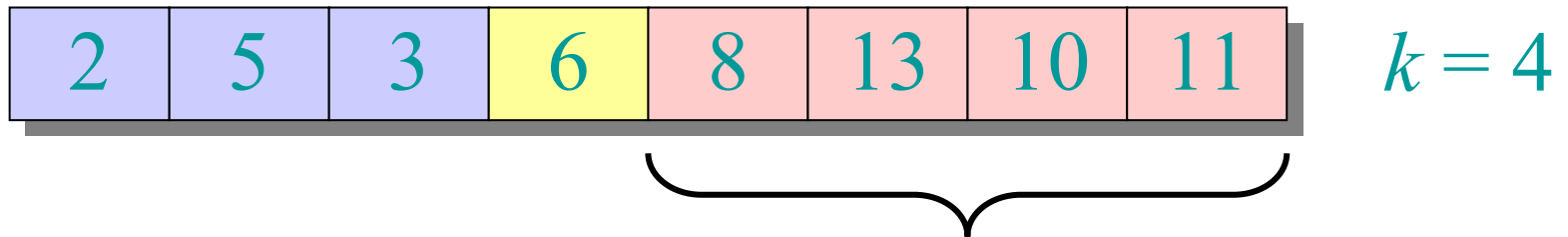
# Example

Select the  $i = 7$ th smallest:

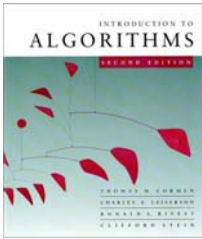


*pivot*

Partition:



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



# Intuition for analysis

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

**Lucky:**

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

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

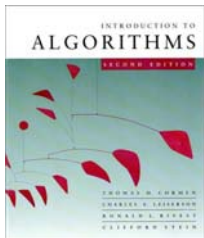
CASE 3

**Unlucky:**

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

arithmetic series

***Worse than sorting!***



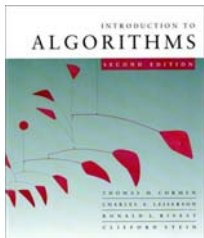
# Analysis of expected time

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*

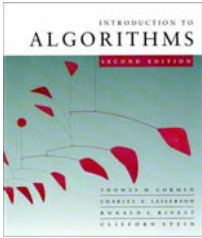
$$X_k = \begin{cases} 1 & \text{if PARTITION generates a } k : n-k-1 \text{ split,} \\ 0 & \text{otherwise.} \end{cases}$$



# Analysis (continued)

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

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

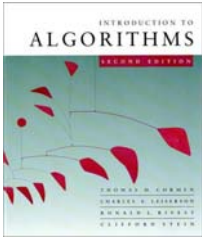


# Calculating expectation

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

Take expectations of both sides.

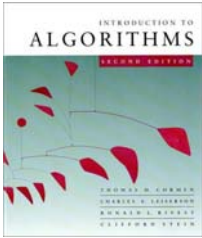




# Calculating expectation

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

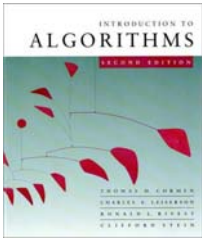
Linearity of expectation.



# Calculating expectation

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

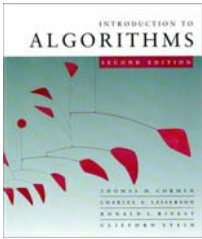
Independence of  $X_k$  from other random choices.



# Calculating expectation

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

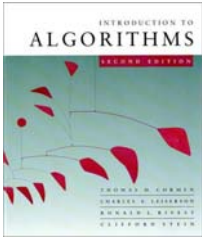
Linearity of expectation;  $E[X_k] = 1/n$ .



# Calculating expectation

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

Upper terms appear twice.



# Hairy recurrence

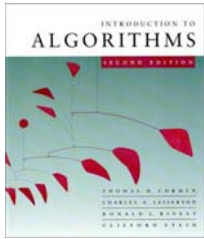
(But not quite as hairy as the quicksort one.)

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

**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.

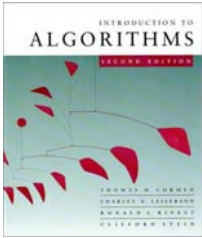
**Use fact:**  $\sum_{k=\lfloor n/2 \rfloor}^{n-1} k \leq \frac{3}{8}n^2$  (exercise).



# Substitution method

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

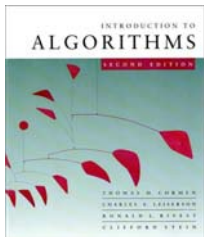
Substitute inductive hypothesis.



# Substitution method

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

Use fact.

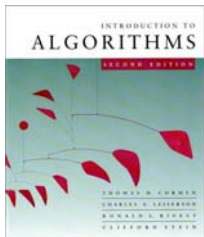


# Substitution method

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

Express as *desired – residual*.

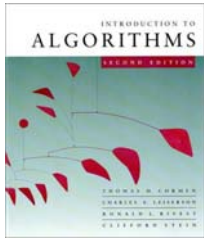




# Substitution method

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

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



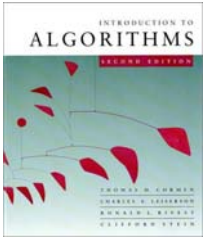
# Summary of randomized order-statistic selection

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

**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.

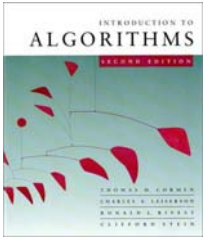


# Worst-case linear-time order statistics

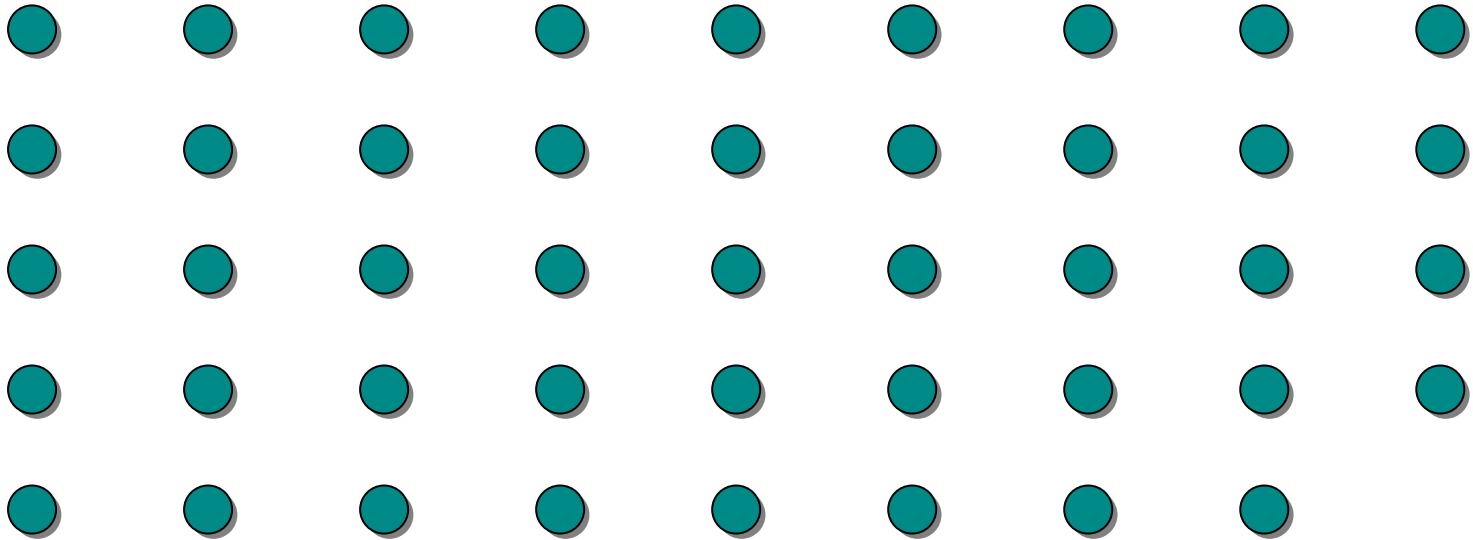
SELECT( $i, n$ )

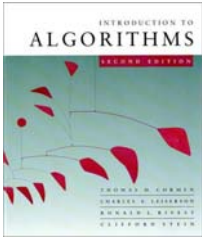
1. Divide the  $n$  elements into groups of 5. Find the median of each 5-element group by rote.
2. Recursively SELECT the median  $x$  of the  $\lfloor n/5 \rfloor$  group medians to be the pivot.
3. Partition around the pivot  $x$ . Let  $k = \text{rank}(x)$ .
4. **if**  $i = k$  **then return**  $x$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

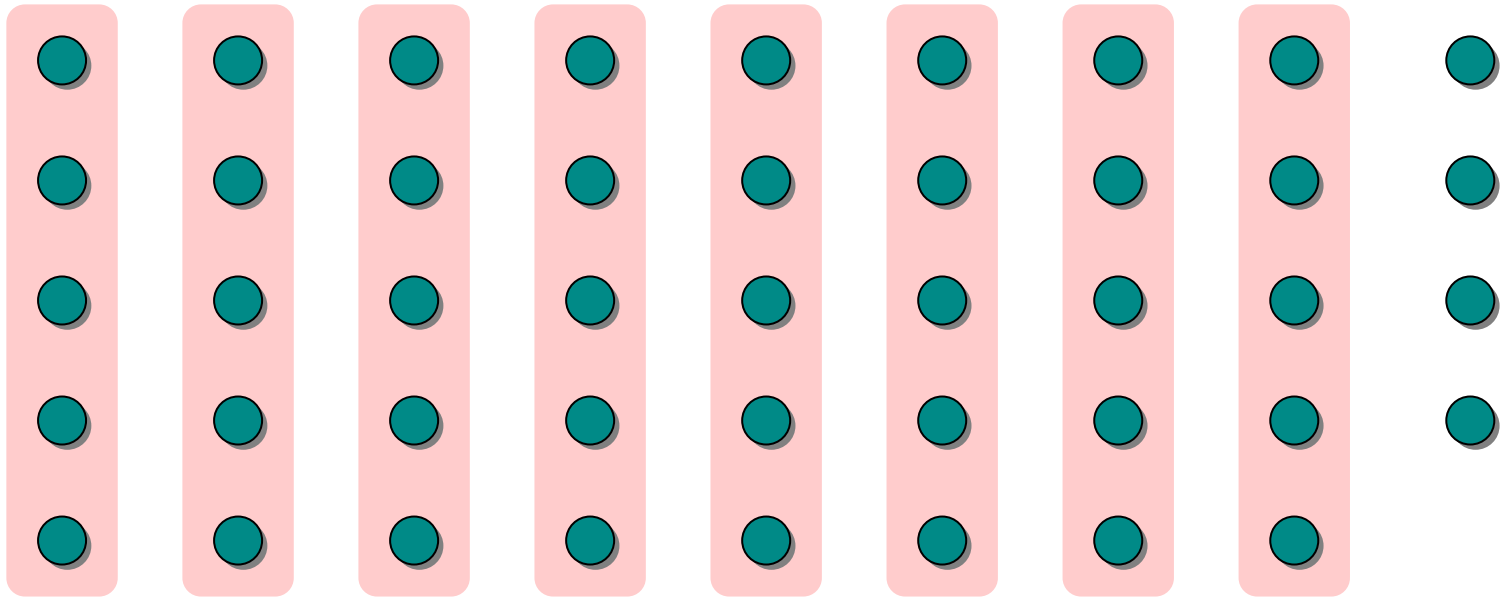


# Choosing the pivot

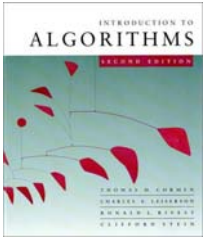




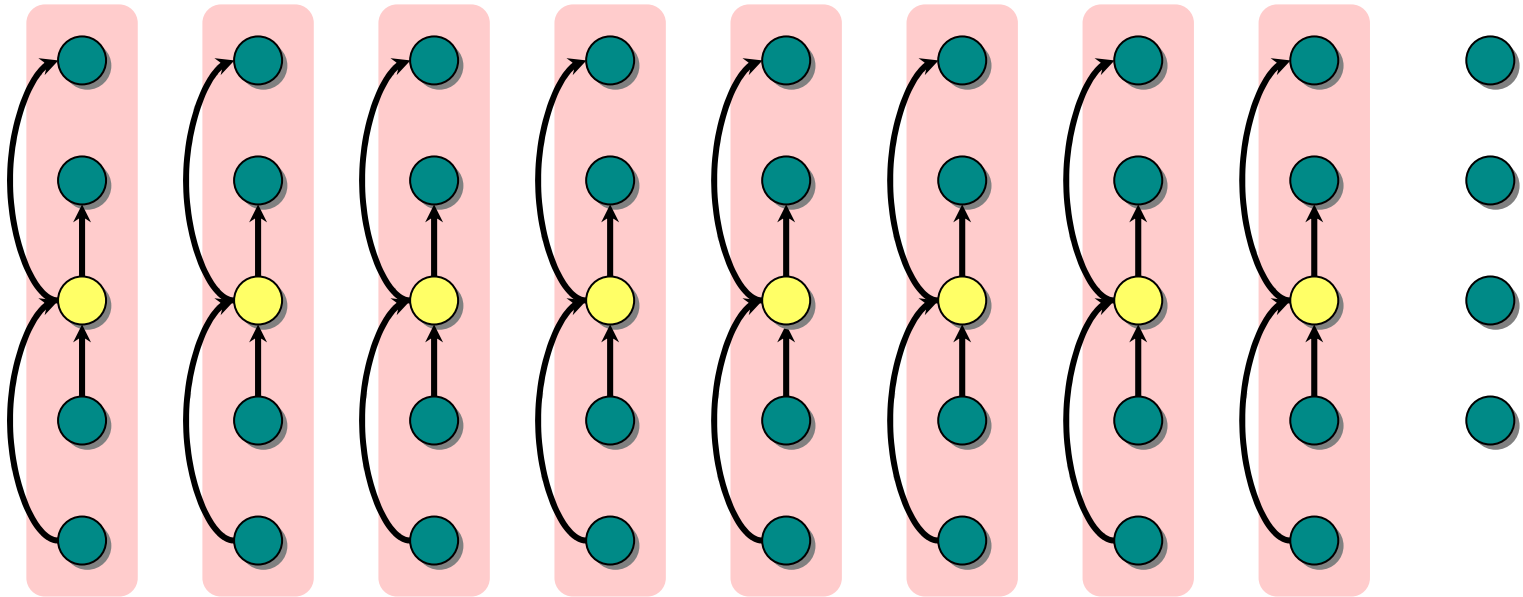
# Choosing the pivot



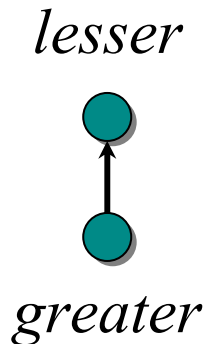
1. Divide the  $n$  elements into groups of 5.

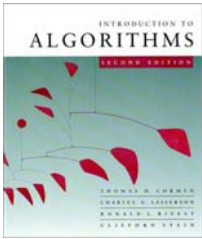


# Choosing the pivot

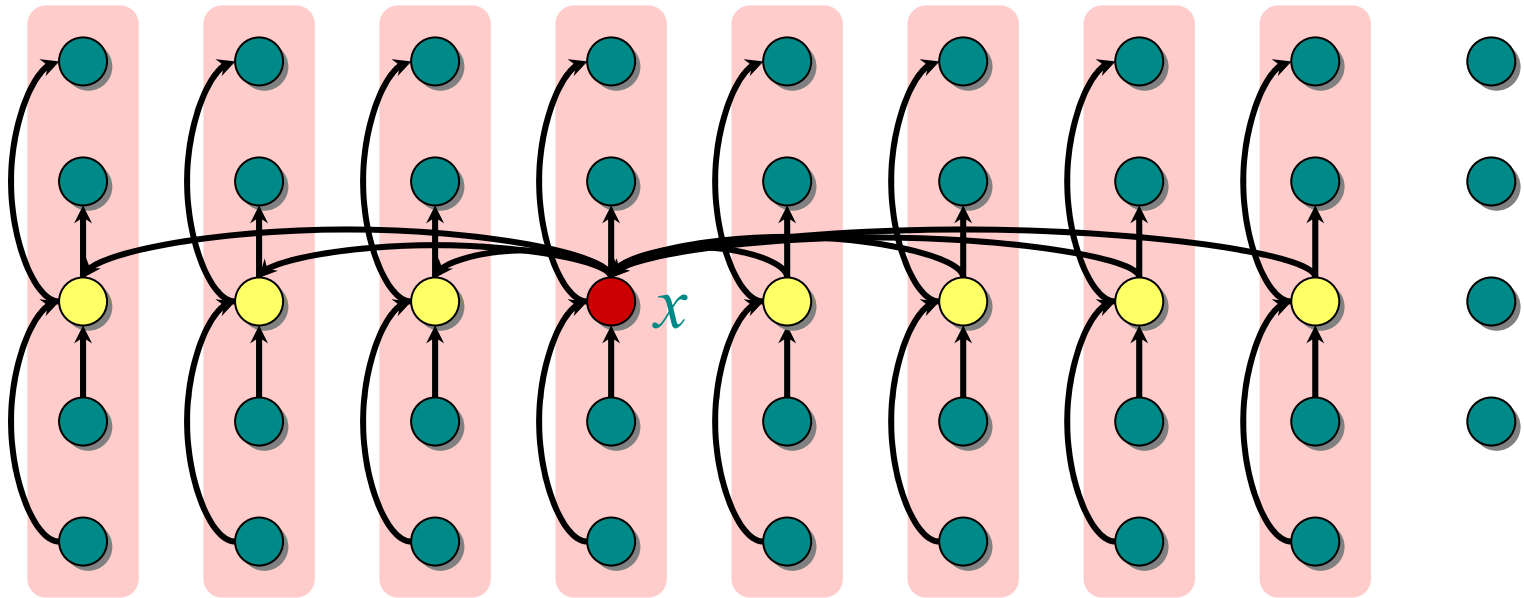


1. Divide the  $n$  elements into groups of 5. Find the median of each 5-element group by rote.

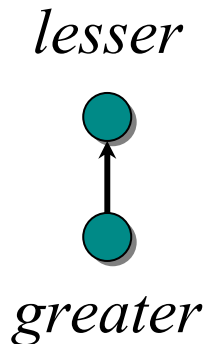


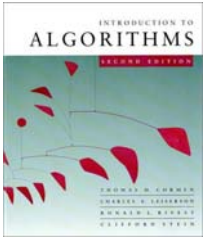


# Choosing the pivot

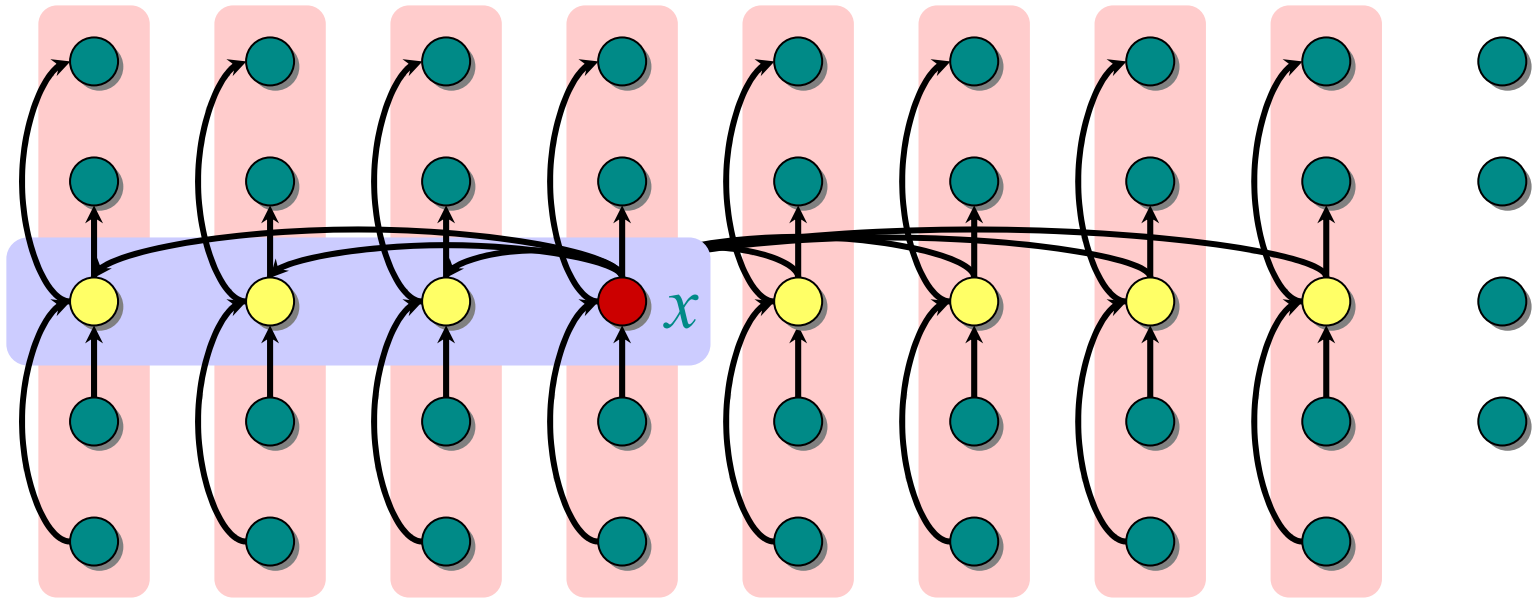


1. Divide the  $n$  elements into groups of 5. Find the median of each 5-element group by rote.
2. Recursively SELECT the median  $x$  of the  $\lfloor n/5 \rfloor$  group medians to be the pivot.



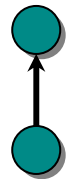


# Analysis



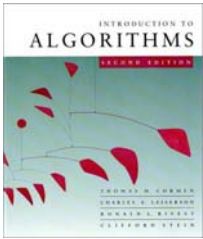
At least half the group medians are  $\leq x$ , which is at least  $\lfloor \lfloor n/5 \rfloor / 2 \rfloor = \lfloor n/10 \rfloor$  group medians.

*lesser*

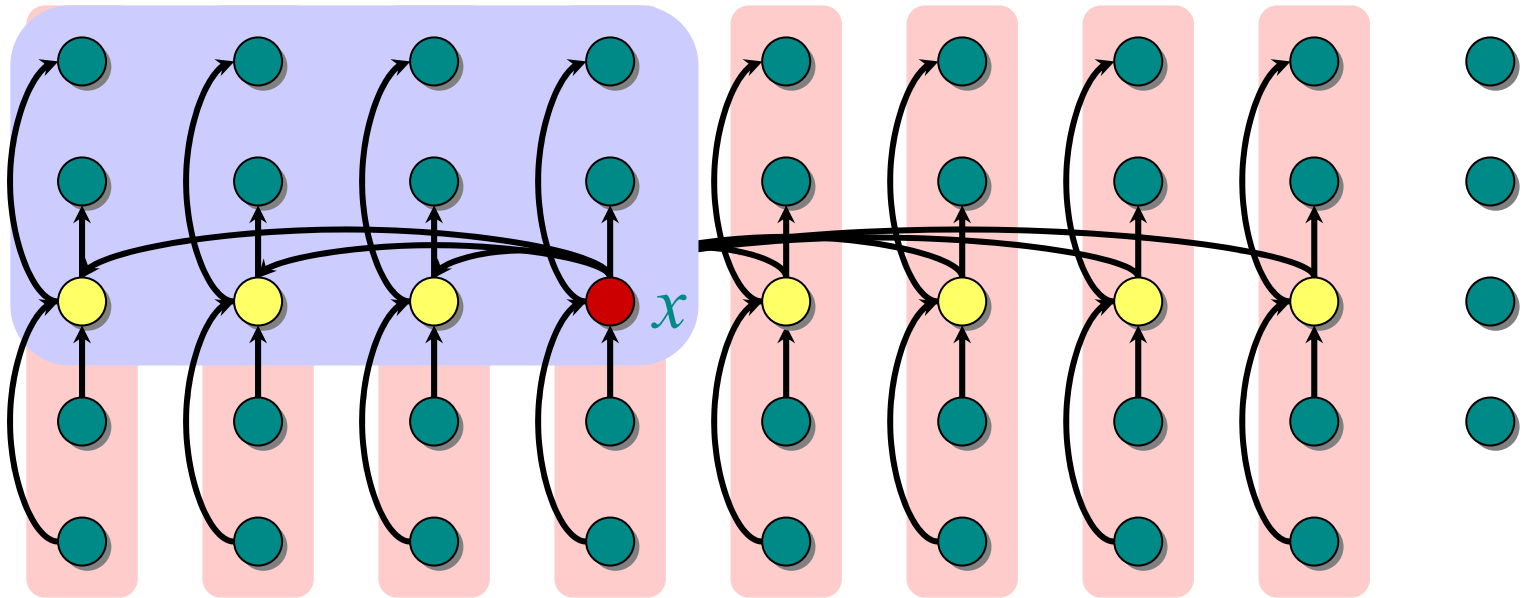


*greater*



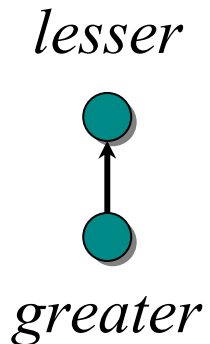


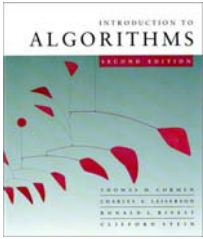
# Analysis (Assume all elements are distinct.)



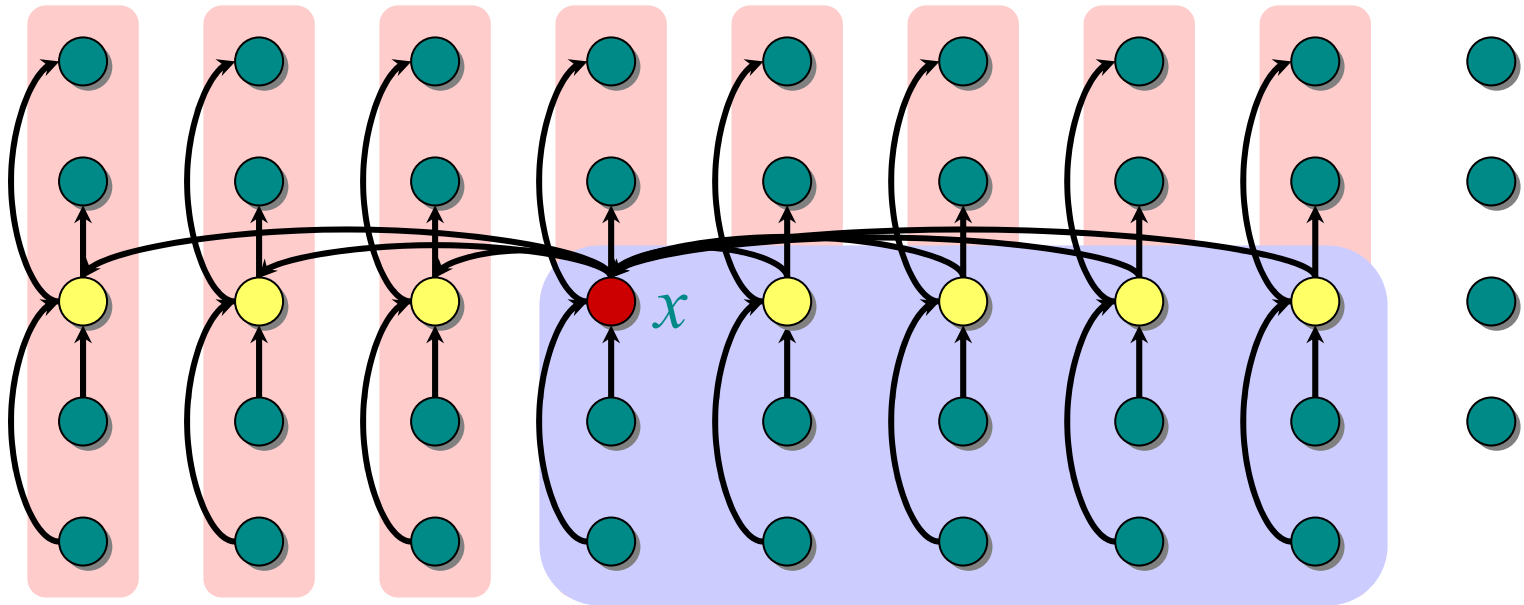
At least half the group medians are  $\leq x$ , which is at least  $\lfloor \lfloor n/5 \rfloor / 2 \rfloor = \lfloor n/10 \rfloor$  group medians.

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





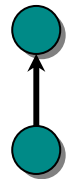
# Analysis (Assume all elements are distinct.)



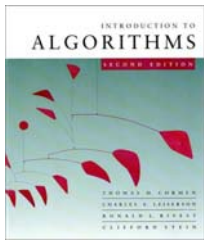
At least half the group medians are  $\leq x$ , which is at least  $\lfloor \lfloor n/5 \rfloor / 2 \rfloor = \lfloor n/10 \rfloor$  group medians.

- 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*

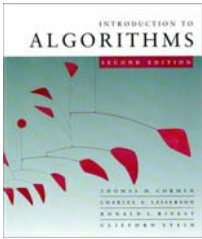


*greater*



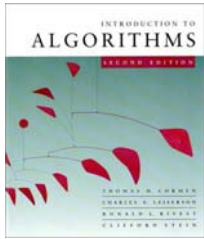
# Minor simplification

- 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  $\leq 3n/4$  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**( $i, n$ )
- 
- $\Theta(n)$       {
- $T(n/5)$       {
- $\Theta(n)$       {
- $T(3n/4)$       {
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



# Solving the recurrence

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

**Substitution:**

$$T(n) \leq cn$$

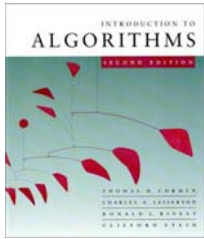
$$T(n) \leq \frac{1}{5}cn + \frac{3}{4}cn + \Theta(n)$$

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

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

$$\leq cn \text{ ,}$$

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?*