



## Lecture ~~7~~: Order Statistics

# ORDER STATISTICS

Select the  $i$ th smallest of  $n$  elements.

- $i = 1$ : *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).

# RANK OF AN ELEMENT

Rank of an element is the position of the element in the sorted array

Given Numbers:

10	13	5	8	3	2	11
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Sorted Array:

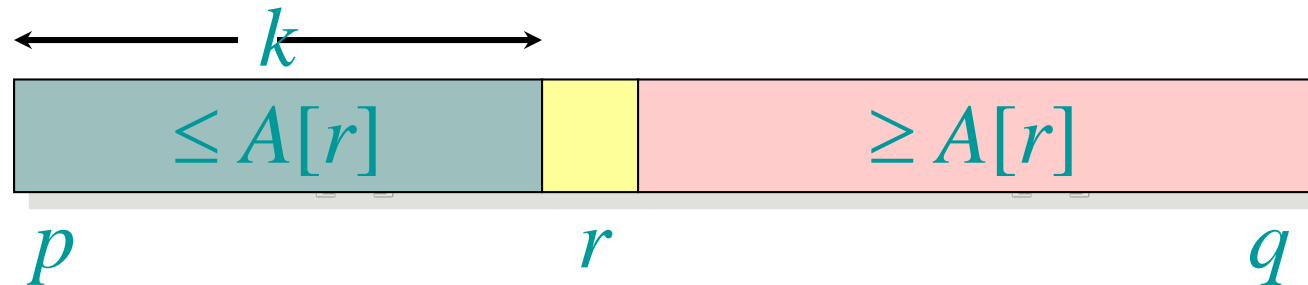
2	3	5	8	10	11	13
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$$\text{rank}(8)=4$$

Finding the  $i$ th smallest of  $n$  elements = Finding an element with *rank*  $i$ .

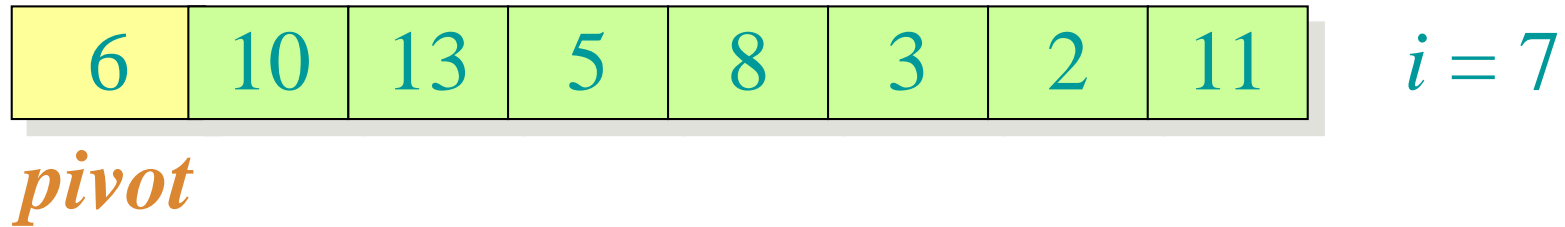
# DIVIDE-AND-CONQUER ALGORITHM

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

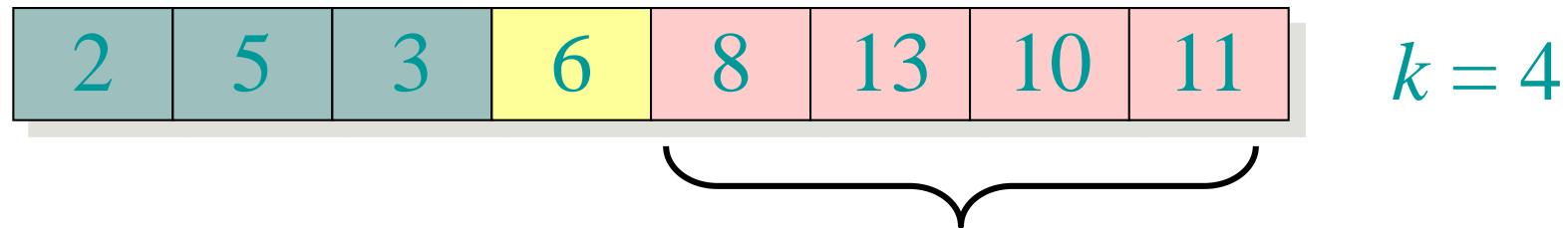


# EXAMPLE

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:**

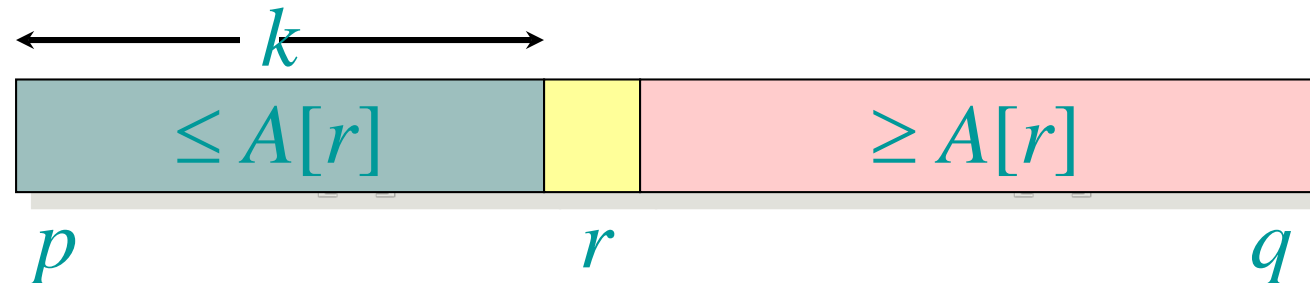
$$\begin{aligned} T(n) &= T(n - 1) + \Theta(n) \\ &= \Theta(n^2) \end{aligned}$$

arithmetic series

*Worse than sorting!*

# 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**( $A, p, q$ )  
   $k \leftarrow r - p + 1$   
  **if**  $i = k$  **then return**  $A[r]$   
  **if**  $i < k$   
    **then return** **RAND-SELECT**( $A, p, r - 1, i$ )  
  **else return** **RAND-SELECT**( $A, r + 1, q, i - k$ )



# 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

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

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.

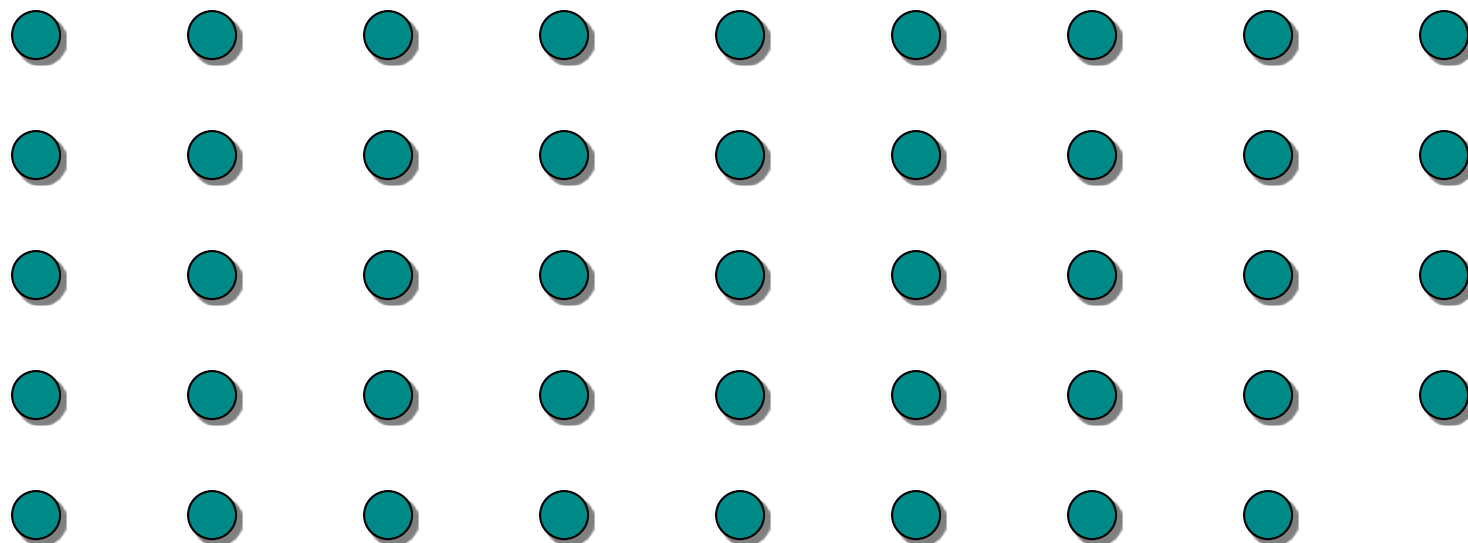
# WORST-CASE LINEAR-TIME ORDER STATISTICS

SELECT( $i, n$ )

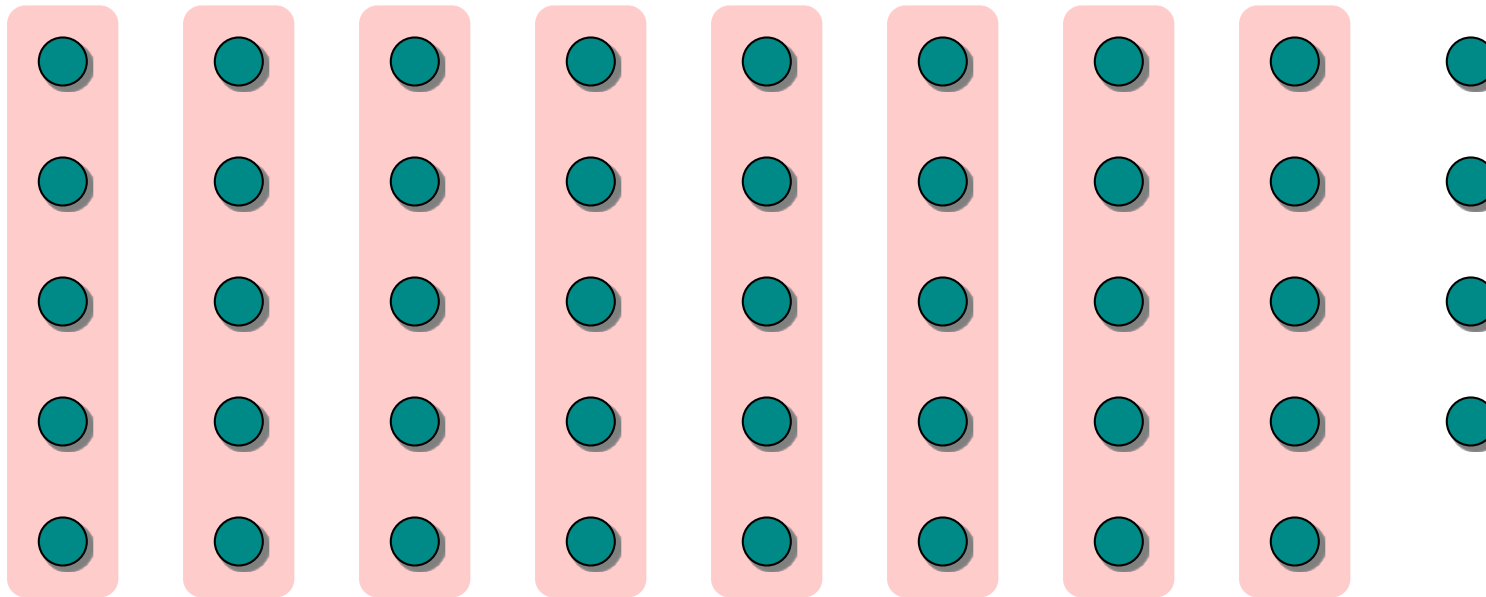
1. Divide the  $n$  elements into groups of 5. Find the median of each 5-element group by rote.
2. Recursively SELECT the median  $x$  of the  $\lfloor n/5 \rfloor$  group medians to be the pivot.
3. Partition around the pivot  $x$ . Let  $k = \text{rank}(x)$ .
4. **if**  $i = k$  **then return**  $x$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

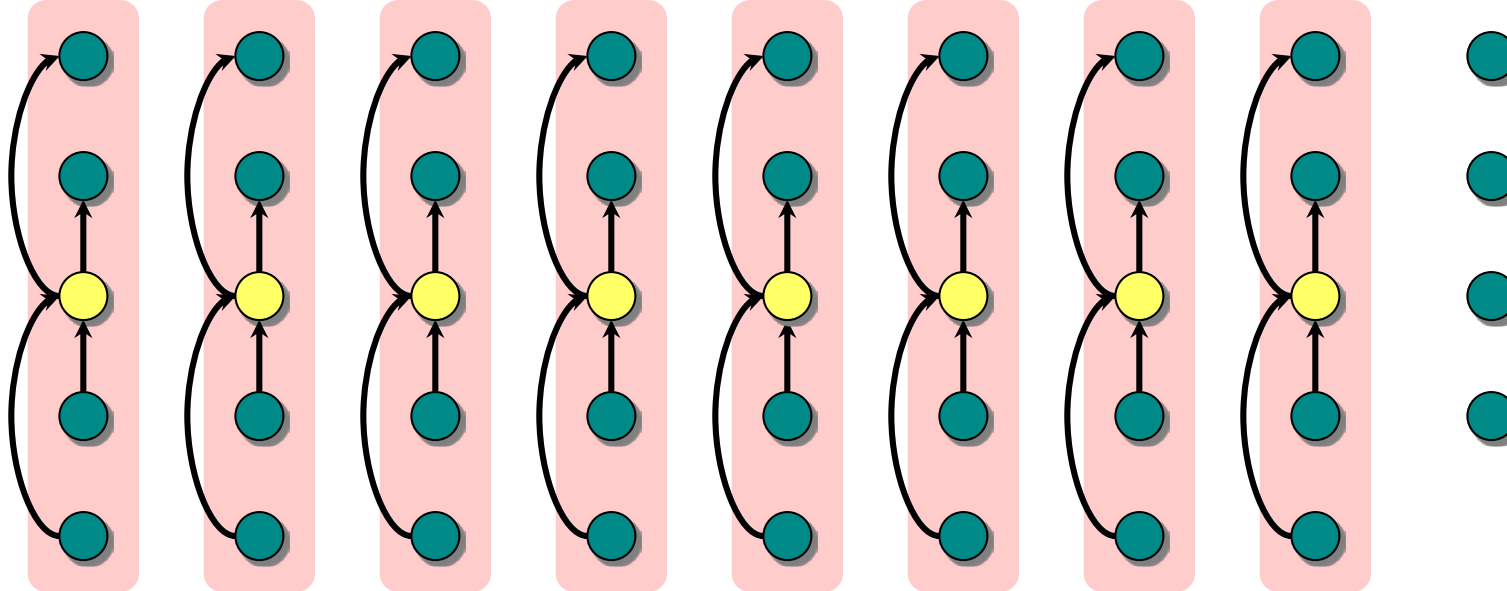


# CHOOSING THE PIVOT

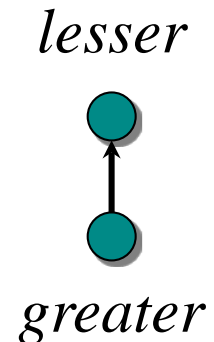


1. Divide the  $n$  elements into groups of 5.

# CHOOSING THE PIVOT

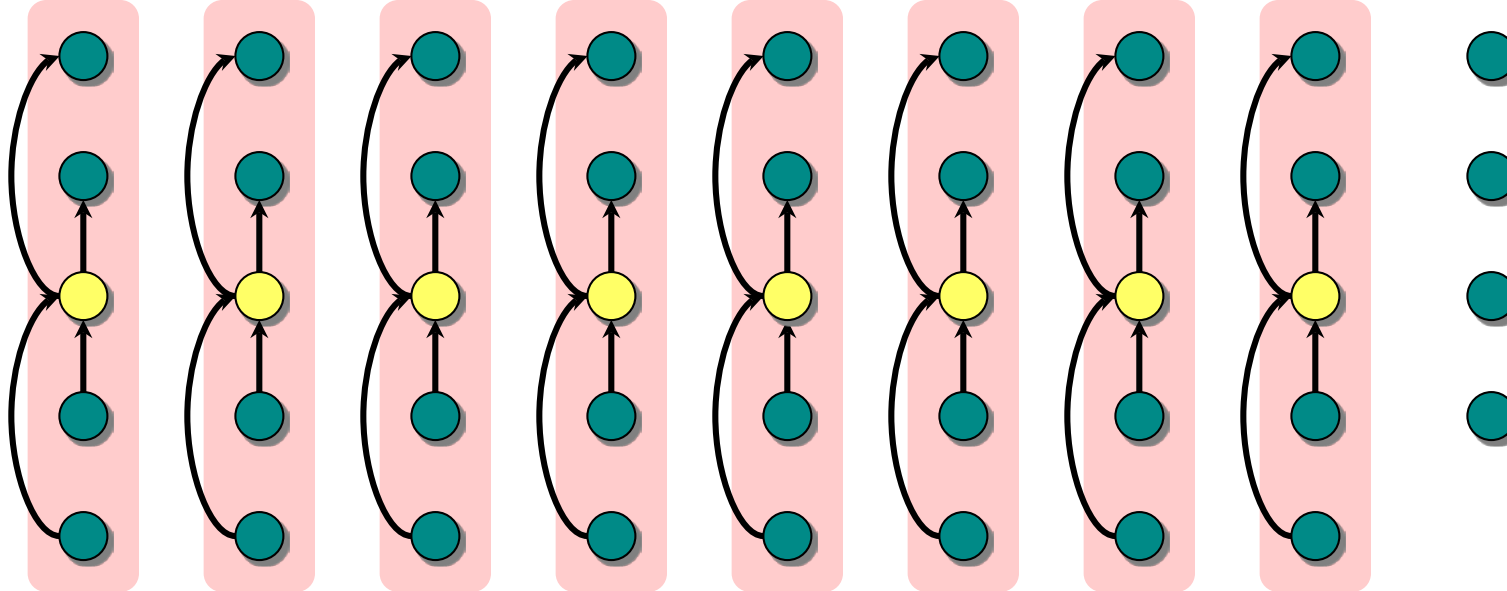


1. Divide the  $n$  elements into groups of 5. Find the median of each 5-element group by rote.

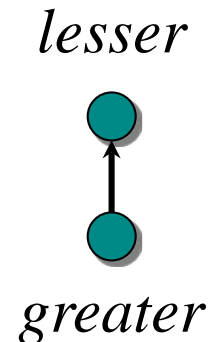




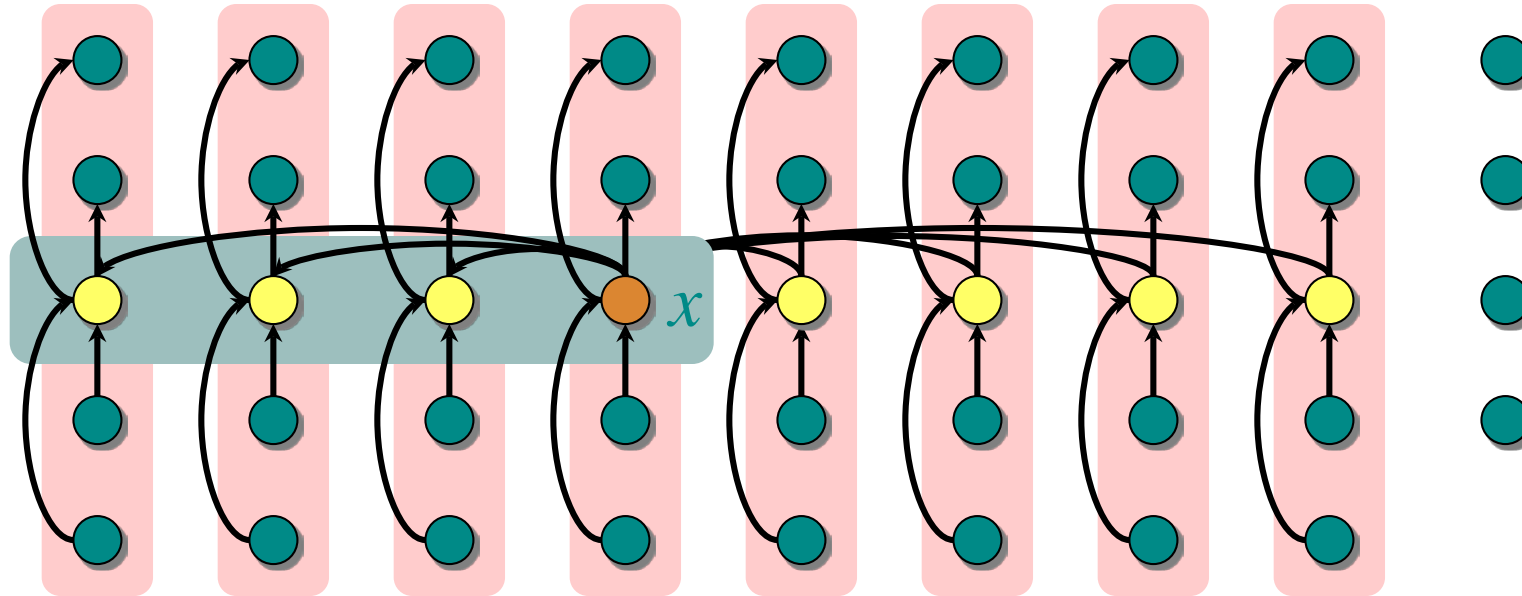
# CHOOSING THE PIVOT




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.



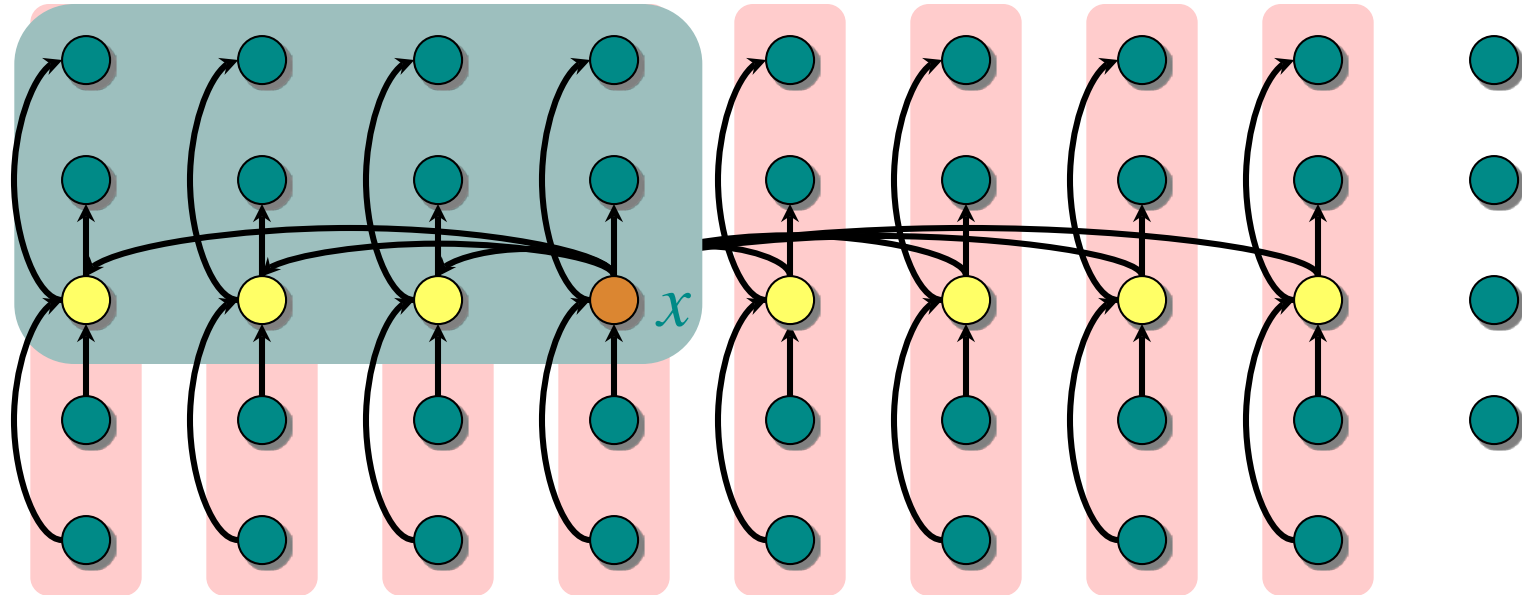
# ANALYSIS



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.

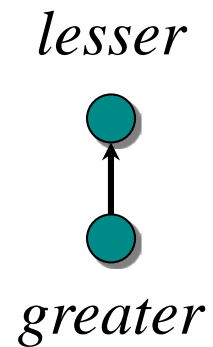
*lesser*  
  
*greater*

# ANALYSIS

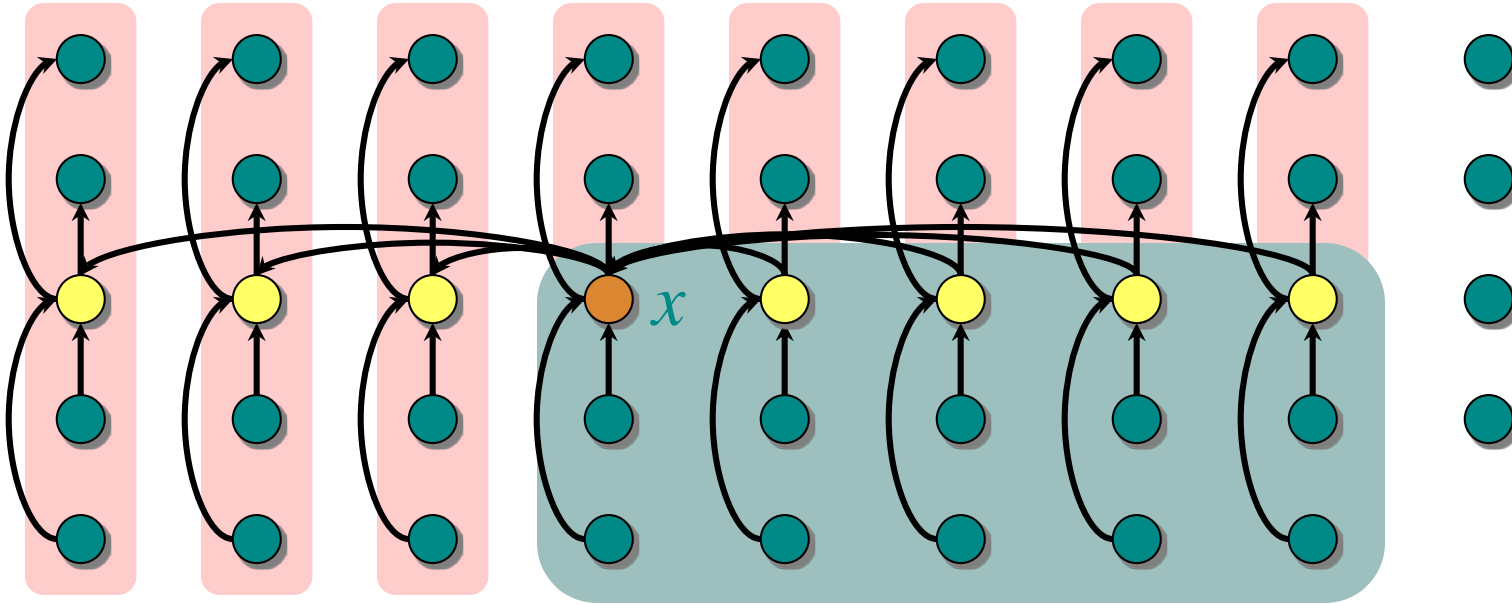


At least half the group medians are  $\leq x$ , which is at least  $\lfloor \lfloor n/5 \rfloor / 2 \rfloor = \lfloor n/10 \rfloor$  group medians.

- 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*

# DEVELOPING THE RECURRENCE

$T(n)$	SELECT( $i, n$ )
$\Theta(n)$	1. Divide the $n$ elements into groups of 5. Find the median of each 5-element group by rote.
$T(n/5)$	2. Recursively SELECT the median $x$ of the $\lfloor n/5 \rfloor$ group medians to be the pivot.
$\Theta(n)$	3. Partition around the pivot $x$ . Let $k = \text{rank}(x)$ .
$T(7n/10)$	4. <b>if</b> $i = k$ <b>then return</b> $x$ <b>elseif</b> $i < k$ <b>then</b> recursively SELECT the $i$ th smallest element in the lower part <b>else</b> recursively SELECT the $(i-k)$ th smallest element in the upper part

# SOLVING THE RECURRENCE

$$T(n) = T(n/5) + T(7n/10) + \Theta(n)$$

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**Substitution:**

$$\begin{aligned} T(n) \leq cn \quad T(n) &\leq cn/5 + 7cn/10 + \Theta(n) \\ &= 9cn/10 + \Theta(n) \\ &= cn - \left[ cn/10 - \Theta(n) \right] \\ &\leq cn \end{aligned}$$

,  
if  $c$  is chosen large enough to handle both the  $\Theta(n)$  and the initial conditions.  $7n/10$

# CONCLUSIONS

- Since the work at each level of recursion is a constant fraction ( $9/10$ ) 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.