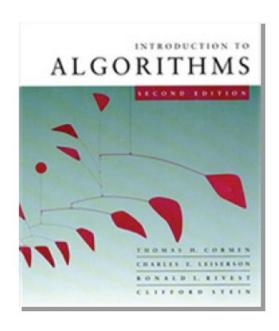
Introduction to Algorithms 6.046J/18.401J



LECTURE 3

Order Statistics

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

Based on slides by Prof. Erik Demaine



Order statistics

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

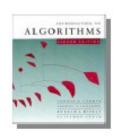
- •i = 1: minimum;
- •i = n: maximum;
- • $i = \lfloor (n+1)/2 \rfloor$ or $\lceil (n+1)/2 \rceil$: median.

Naive algorithm: Sort and index *i*th element.

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

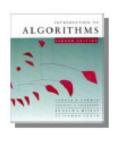
= $\Theta(nlg \ n)$,

using merge sort or heapsort (not quicksort).



Randomized divide-and-conquer algorithm

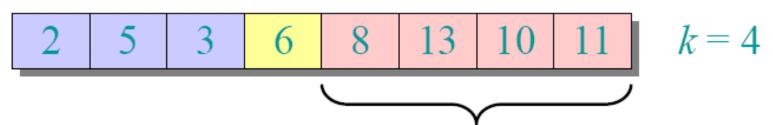
```
RAND-SELECT (A, p, q, i) \triangleright i th smallest of A[p ... q]
 if p = q then return A[p]
 r \leftarrow \text{RAND-PARTITION}(A, p, q)
                  \triangleright k = \operatorname{rank}(A[r])
 k \leftarrow r - p + 1
 if i = k then return A[r]
 if i < k
     then return RAND-SELECT (A,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



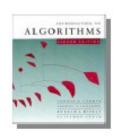
Example

Select the i = 7th smallest:

Partition:



Select the 7 - 4 = 3 rd smallest recursively.



Intuition for analysis

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

Lucky:

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

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

Case 3

Unlucky:

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

= $\Theta(n^2)$

arithmetic series



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-SELECTON an input of size n, assuming random numbers are independent.

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

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

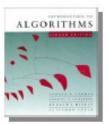


Analysis (continued)

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

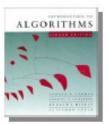
$$T(n) = \begin{cases} T(\max\{0, n-1\}) + \Theta(n) & \text{if } 0: n-1 \text{ split,} \\ T(\max\{1, n-2\}) + \Theta(n) & \text{if } 1: n-2 \text{ split,} \\ \vdots & & \\ T(\max\{n-1, 0\}) + \Theta(n) & \text{if } n-1: 0 \text{ split,} \end{cases}$$

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



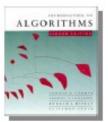
$$E[T(n)] = E\left[\sum_{k=0}^{n-1} X_k (T(\max\{k, n-k-1\}) + \Theta(n))\right]$$

Take expectations of both sides.



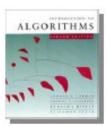
$$\begin{split} E[T(n)] &= E\bigg[\sum_{k=0}^{n-1} X_k \big(T(\max\{k, n-k-1\}) + \Theta(n) \big) \bigg] \\ &= \sum_{k=0}^{n-1} E\big[X_k \big(T(\max\{k, n-k-1\}) + \Theta(n) \big) \big] \end{split}$$

Linearity of expectation.



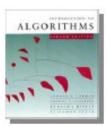
$$\begin{split} E[T(n)] &= E\bigg[\sum_{k=0}^{n-1} X_k \big(T(\max\{k, n-k-1\}) + \Theta(n) \big) \bigg] \\ &= \sum_{k=0}^{n-1} E\big[X_k \big(T(\max\{k, n-k-1\}) + \Theta(n) \big) \big] \\ &= \sum_{k=0}^{n-1} E\big[X_k \big] \cdot E\big[T(\max\{k, n-k-1\}) + \Theta(n) \big] \end{split}$$

Independence of X_k from other random choices



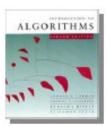
$$\begin{split} E[T(n)] &= E \Bigg[\sum_{k=0}^{n-1} X_k \big(T(\max\{k, n-k-1\}) + \Theta(n) \big) \Bigg] \\ &= \sum_{k=0}^{n-1} E \big[X_k \big(T(\max\{k, n-k-1\}) + \Theta(n) \big) \big] \\ &= \sum_{k=0}^{n-1} E \big[X_k \big] \cdot E \big[T(\max\{k, n-k-1\}) + \Theta(n) \big] \\ &= \frac{1}{n} \sum_{k=0}^{n-1}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

$$E[X_k] = 1/n.$$



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

This part is $\Theta(n)$



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

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

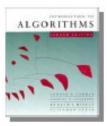
$$= \sum_{k=0}^{n-1} E\left[X_k\right] \cdot E\left[T(\max\{k, n-k-1\}) + \Theta(n)\right]$$

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

$$\leq \frac{2}{n} \sum_{k=\lfloor n/2 \rfloor}^{n-1} E\left[T(k)\right] + \Theta(n) \quad \text{Upper terms appear twice.}$$

$$E.g. \ k = 1 \text{ vs. } k = n - 2$$

$$max(1, n-2) = max(n-2, 1)$$



Hairy recurrence

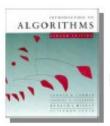
(But not quite as hairy as the quicksort one.)

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

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

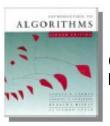
• The constant c can be chosen large enough so that $E/T(n)/\leq cn$ for the base cases.

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



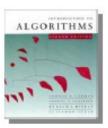
$$E[T(n)] \le \frac{2}{n} \sum_{k=\lfloor n/2 \rfloor}^{n-1} ck + \Theta(n)$$

Substitute inductive hypothesis.



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

Use fact.

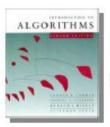


$$E[T(n)] \le \frac{2}{n} \sum_{k=\lfloor n/2 \rfloor}^{n-1} ck + \Theta(n)$$

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

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

Express as desired-residual.



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

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

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

$$\le cn,$$

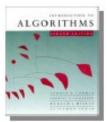
if c is chosen large enough so that cn/4 dominates the $\Theta(n)$.



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

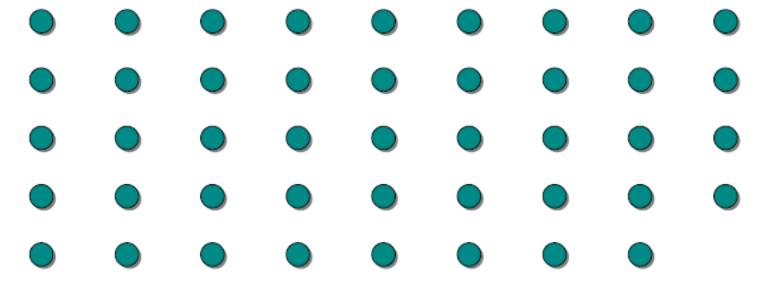
SELECT(i, n)

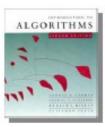
- 1. Divide the *n* elements into groups of *5*. Find the median of each *5*-element group by rote.
- 2. Recursively SELECT the median x of the $\lfloor n/5 \rfloor$ group medians to be the pivot.
- 3. Partition around the pivot x. Let k = rank(x).
- 4. If i = k then return x else if i < k

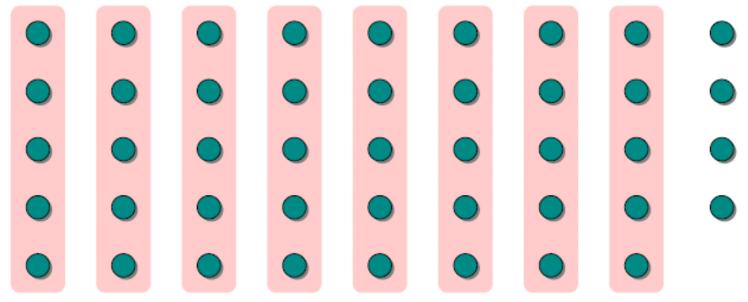
then recursively SELECT the *i* th smallest element in the lower part else recursively SELECT the (*i-k*) th smallest element in the upper part

Same as RAND-SELECT

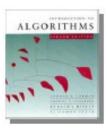


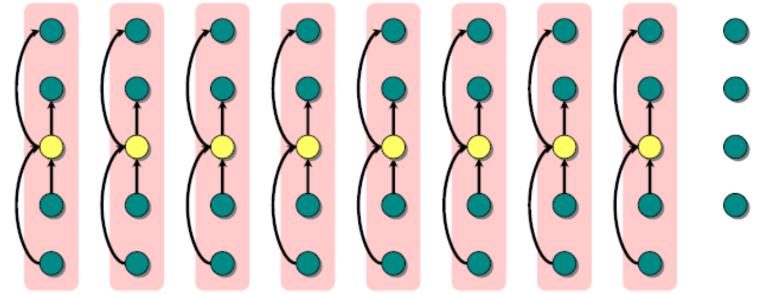




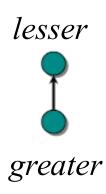


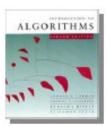
1. Divide the *n* elements into groups of 5.

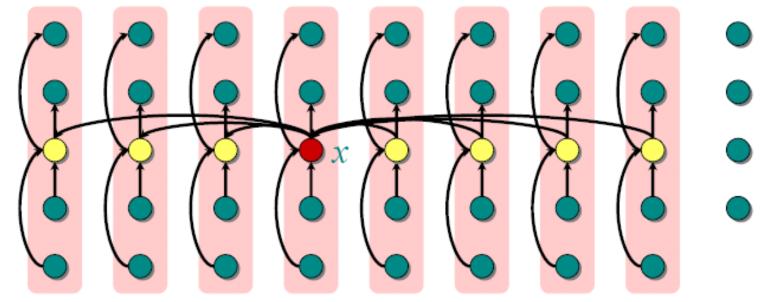




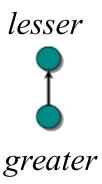
1. Divide the *n* elements into groups of *5*. Find the median of each *5*-element group by rote.





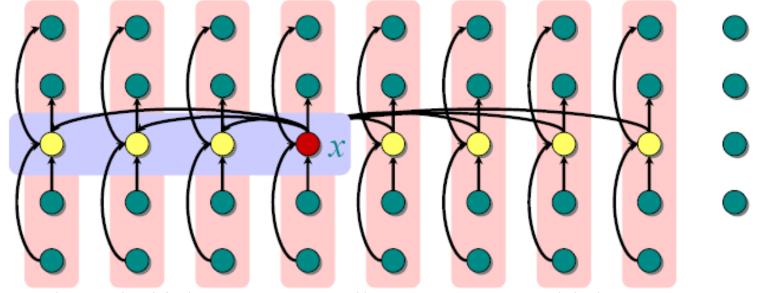


- 1. Divide the *n* elements into groups of *5*. Find the median of each *5*-element group by rote.
- 2. Recursively SELECT the median x of the $\lfloor n/5 \rfloor$ group medians to be the pivot.

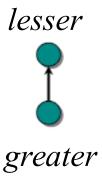




Analysis

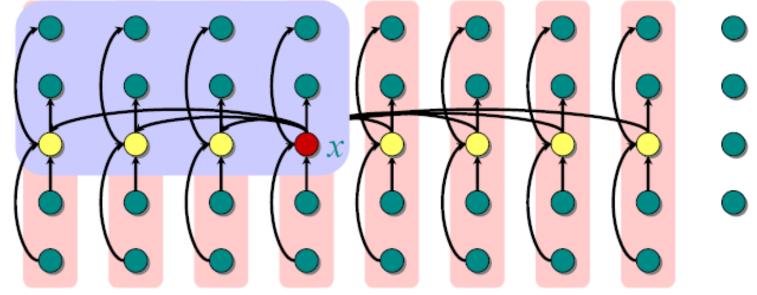


At least half the group medians are $\leq x$, which is at least $\lfloor n/5 \rfloor /2 \rfloor = \lfloor n/10 \rfloor$ group medians.





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.

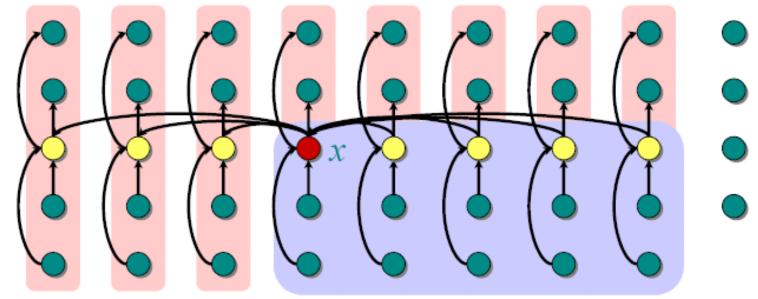
• Therefore, at least $3\lfloor n/10 \rfloor$ elements are $\leq x$.

lesser

greater



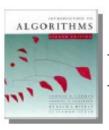
Analysis (Assume all elements are distinct.)



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.

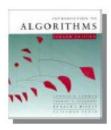
- Therefore, at least $3 \mid n/10 \mid$ elements are $\leq x$.
- Similarly, at least $3\lfloor n/10 \rfloor$ elements are $\geq x$.

lesser
greater



Minor simplification

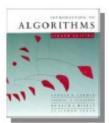
- For $n \ge 50$, we have $3 \lfloor n/10 \rfloor \ge n/4$.
- Therefore, for $n \ge 50$ the recursive call to SELECT in Step 4 is executed recursively on $\le 3n/4$ elements.
- Thus, the recurrence for running time can assume that Step 4 takes time T(3n/4) in the worst case.
- For n < 50, we know that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



Developing the recurrence

```
SELECT(i, n)
  T(n)
            1. Divide the n elements into groups of 5. Find the median of each 5-element group by rote.
the median of each 3-element group by rote.

T(n/5) = \begin{cases}
2. & \text{Recursively SELECT the median } x \text{ of the } n/5 \\
& \text{group medians to be the pivot.}
\end{cases}
               3. Partition around the pivot x. Let k = rank(x).
               4. If i = k then return x else if 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:

$$T(n) \le cn$$

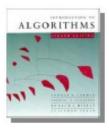
$$T(n) \le \frac{1}{5}cn + \frac{3}{4}cn + \Theta(n)$$

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

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

$$\le cn$$

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?