# Lecture: Analysis of Algorithms (CS583-004) 

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Spring 2019
(1) Outline of Today's Class
(2) Order Statistics

- Selection of Order Statistics in Expected Linear Time - Randomized Divide and Conquer
- Selection of Order Statistics in Worst-case Linear Time
- Median of Medians
- Analysis of Worst-case Running Time
- Order Statistics: Conclusions


## Order Statistics

## Some Order Statistics We Know

Select the $i^{\text {th }}$ smallest of $n$ elements (the element with rank $i$ ):

- $i=1$ : minimum
- $i=n$ : maximum
- $i=(n+1) / 2$ : median


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Naive algorithm: Sort and index $i^{\text {th }}$ element.

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## Randomized Divide and Conquer Algorithm

RAND-SELECT(array A, p, q, i) $\triangleright i^{\text {th }}$ smallest of $A[p \ldots q]$
1: if $p=q$ then
2: return $A[p]$
3: $r \leftarrow \operatorname{RAND}-\mathrm{PARTITION}(A, p, q)$
4: $k \leftarrow r-p+1 \quad \triangleright k=\operatorname{rank}(\mathrm{A}[\mathrm{r}])$
5: if $i=k$ then
6: return $A[r]$
7: if $i<k$ then
8: return RAND-SELECT $(A, p, r-1, i)$
9: else return RAND-SELECT $(A, r+1, q, i-k)$

$p$
r

## Randomized Select: Trace

Select the $i=7^{\text {th }}$ smallest element from the array below:

| 6 | 10 | 13 | 5 | 8 | 3 | 2 | 11 |
| :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- |

## pivot

## Partition:



Now select the 7-4 = $3^{\text {rd }}$ smallest element recursively.

## Randomized Select: Running Time Analysis

- Analysis follows closely that of quicksort
- For simplicity, we will assume that all elements are distinct
- We will first gain intuition through lucky/unlucky scenarios

Lucky: [assume a 1:9 partition after RAND-PARTITION]

$$
\begin{aligned}
T(n) & =T(9 n / 10)+\theta(n) & & n^{\log _{10 / 9}(1)}=n^{0}=1 \\
& =\theta(n) & & \text { CASE } 3 \text { of master theorem }
\end{aligned}
$$

Unlucky: [assume one side of the partitioned array is empty]

$$
\begin{aligned}
T(n) & =T(n-1)+\theta(n) & & \text { arithmetic series } \\
& =\theta\left(n^{2}\right) & & \text { worse than sorting!!! }
\end{aligned}
$$

## Randomized Select: Analysis of Expected Time

- Analysis similar to randomized quicksort
- Let $T(n)$ be the random variable for the running time of RAND-SELECT on an input of size $n$, assuming random numbers are independent
- To obtain upper bound, assume the $i^{\text {th }}$ smallest element always falls on the larger side of the partition:

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

- Summing up we have:

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

## Randomized Select: Those pesky expectations...

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\begin{array}{r}
E[T(n)]=\frac{1}{n} \sum_{k=0}^{n-1} E[T(\max \{k, n-k-1\})]+\theta(n) \\
=\frac{\operatorname{get} \theta(n) \text { outside }}{n} \sum_{k=\lfloor n / 2\rfloor}^{n-1} E[T(k)]+\theta(n) \\
\text { upper terms appear twice }
\end{array}
$$

Outline of Today's Class

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Prove: $E[T(n)] \leq c \cdot n$ for $c>0$ ( $c$ large enough for base cases)

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$$
\sum_{k=\lfloor n / 2\rfloor}^{n-1} k \leq \frac{3}{8} n^{2} \text { (exercise: show it) }
$$

So: $E[T(n)] \leq c n-\left(\frac{c n}{4}-\theta(n)\right) \leq c n$ if $\frac{c n}{4}-\theta(n) \geq 0$

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Easy to find a large value of $c$ such that $\frac{c n}{4}$ dominates $\theta(n)$

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## Randomized Select: Summary

- Works fast in the average case: linear expected time
- Very simple and fast algorithm in practice
- But, worst-case behavior is $\theta\left(n^{2}\right)$
- Question: Is there an algorithm that runs in linear time even in the worst case?
- Answer: Yes - in 1973, Blum, Floyd, Pratt, and Rivest designed such an algorithm
- Basic Idea: Generate good pivots recursively to guarantee a good split


## 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. Let $k=\operatorname{rank}(x)$
4: if $i=k$ then
5: return $x$
6: if $i<k$ then
7: recursively SELECT $i^{\text {th }}$ smallest element in lower part
8: if $i>k$ then
9: recursively SELECT $(i-k)^{\text {th }}$ smallest element in upper part
Note: lines 3.-9. are the same as in RAND-SELECT

Outline of Today's Class
Order Statistics

## SELECT: Choosing the Pivot



Here is the input: $n$ elements.

## SELECT: Choosing the Pivot


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

## SELECT: Choosing the Pivot


lesser
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## Select: Running Time Analysis


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

## Select: Running Time Analysis


lesser

greater

## Select: Running Time Analysis

 is at least $\lfloor\lfloor n / 5\rfloor / 2\rfloor=\lfloor n / 10\rfloor$ group medians.

- If we assume that all elements are distinct, then there are $3\lfloor n / 10\rfloor$ elements $\leq x$.
- Similarly, at least $3\lfloor n / 10$ 」 elements are $\geq x$

greater


## Select: Running Time Analysis

- For $n \geq 50$, we have $3\lfloor n / 10\rfloor \geq n / 4$. So, the call to SELECT in lines 4 and on is executed recursively on at most $3 n / 4$ elements
- The recurrence for the running time can assume that lines 4 and on takes $T(3 n / 4)$ in the worst case
- For $n<50$, we know that the worst-case time is $T(n) \in \theta(1)$

The recurrence is: $T(n)=T(n / 5)+\theta(n)+T(3 n / 4)$

## Breakdown:

- Line 1: $\theta(n)$
- Line 2: $T(n / 5)$
- Line 3: $\theta(n)$


## Substitution:

$$
\begin{aligned}
T(n) & \leq \frac{1}{5} c \cdot n+\frac{3}{4} c \cdot n+\theta(n) \\
& =\frac{9}{20} c \cdot n+\theta(n) \\
& =c \cdot n-\left(\frac{1}{20} c \cdot n-\theta(n)\right) \\
& \leq c \cdot n
\end{aligned}
$$

- Lines $\geq$ 4: $T(3 n / 4)$
if $c$ is large enough to dominate $\theta(n)$


## Order Statistics: 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 and simpler to implement
- Exercise: Why not divide into groups of 3 ?

