### VE281

Data Structures and Algorithms

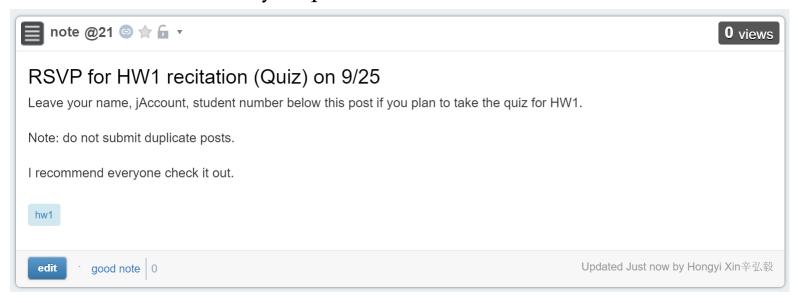
#### **Linear Time Selection**

#### **Learning Objective:**

- Understand randomized selection algorithm
- Understand deterministic selection algorithm
- Know how to analyze their runtime complexity

#### Announcement

- RC class this Friday 8-10PM
  - Post on Piazza if you plan to come to the RC class



• Testing VM will be released tomorrow

### Recap

- What is the complexity of radix sort for decimal numbers?
  - Max digits: K
  - Number of elements: N

# Recap

• What is the worst-case complexity for quicksort?

### Outline

- Randomized selection algorithm
- Deterministic selection algorithm

#### The Selection Problem

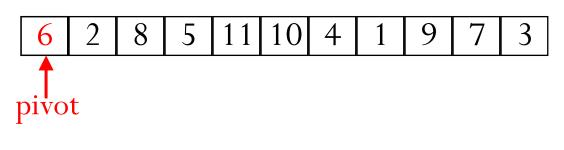
- Input: array A with n distinct numbers and a number i
  - "Distinct" for simplicity
- Output: *i*-th smallest element in the array
  - Assume index starts from 1
- Example: A = (6, 3, 5, 4, 2), i = 3
  - Should return 4
- Special cases
  - i = 1: the smallest item. Runtime: O(n)
  - i = n: the largest item. Runtime: O(n)
  - i = n/2: the median

# Solution: Reduction to Sorting

- Step 1: Do merge sort
- Step 2: output the i-th element of the sorted array
- Time complexity is  $O(n \log n)$
- Can we do better?
  - This essentially asks whether selection is fundamentally easier than sorting
  - Answer: Yes!
  - We will show an O(n) time randomized algorithm by modifying quick sort
  - Also will show an O(n) time deterministic algorithm (However, not as practical as the randomized algorithm)

# Recall: Partitioning in Quick Sort

- Pick a pivot
- Put all elements < pivot to the left of pivot
- Put all elements ≥ pivot to the right of pivot
- Move pivot to its correct place in the array



### Basic Idea

- Suppose we are looking for 6<sup>th</sup> smallest item in an array of length 12. We do partition.
  - Suppose the pivot is at position 4. Then we only need to focus on the sub-array right of the pivot and look for the 2<sup>nd</sup> item in the array
  - Suppose the pivot is at position 8. Then we only need to focus on the sub-array left of the pivot and look for the 6<sup>th</sup> item in the array
  - In both cases, recurse!

#### Randomized Selection

```
Rselect(int A[], int n, int i) {
// find i-th smallest item of array A of size n
  if(n == 1) return A[1];
  Choose pivot p from A uniformly at random;
  Partition A using pivot p;
  Let j be the index of p;
  if(j == i) return p;
  if(j > i) return Rselect(1st part of A, j-1, i);
  else return Rselect(2nd part of A, n-j, i-j);
}
```



### Which Statements Are Correct?

Given a fixed input array, consider the runtime of the randomized selection algorithm to choose the i-th smallest element

- **A.** The runtime depends on the pivot sequence
- **B.** When i = n/2, the worst-case runtime is  $\Theta(n^2)$
- C. When i = n/2, the worst case happens when the pivot sequence is the sorted version of the input array
- **D.** For any given i, the best-case runtime is  $\Theta(1)$

### Short Break - 5 min

• Question Time!

### Average Runtime of Rselect

- Theorem: for every input array of length n, the average runtime of Rselect is O(n)
  - Holds for every input data (no assumption on data)
  - "Average" is over random pivot choices made by the algorithm

# Average Runtime Analysis

- Note: Rselect uses  $\leq cn$  operations outside of recursive call (from partitioning)
- Observation: the length of the array the algorithm works on decreases
- Definition: We say Rselect is in phase j if current array size is between  $(\frac{3}{4})^{j+1}n$  and  $(\frac{3}{4})^{j}n$
- $X_i$  denote the number of recursive calls in phase j
- $runtime \leq \sum_{j} X_{j} \cdot c \cdot (\frac{3}{4})^{j} n$  We need to further get  $E[X_{j}]$

$$E[runtime] \le E\left[\sum_{j} X_{j} \cdot c \cdot \left(\frac{3}{4}\right)^{j} n\right] = cn \sum_{j} \left(\frac{3}{4}\right)^{j} E[X_{j}]$$

# Average Runtime Analysis

- <u>Claim</u>: If Rselect chooses a pivot so that the <u>left sub-array</u>'s size is am, where  $a \in \left[\frac{1}{4}, \frac{3}{4}\right]$  and m is the old length, then the current phase ends
  - Because new sub-array length is at most 75% of the old length
  - "Good pivot"
- What is the probability of  $a \in \left[\frac{1}{4}, \frac{3}{4}\right]$  (i.e., good pivot)?
  - Answer: 0.5
- Claim:  $E[X_j] \le$  Expected number of times you need to get a good pivot
  - Same as the expected number of times you flip a fair coin to get a "head". (Heads: good pivot; tails: bad pivot)

# Coin Flipping Analysis

- Let *N* be the number of coin flips until you get heads
  - N is a geometric random variable:  $P(N = k) = \frac{1}{2^k}, k = 1,2,...$

#flips when 1st is head #flips when 1st is tail

• 
$$E[N] = \frac{1}{2} \cdot 1 + \frac{1}{2} \cdot (1 + E[N]) \Rightarrow E[N] = 2$$

Prob. 1<sup>st</sup> flip is head

Therefore,  $E[X_j] \leq E[N] = 2$ 

Prob. 1st flip is tail

# Average Runtime Analysis

$$E[runtime] \le E\left[\sum_{j} X_{j} \cdot c \cdot \left(\frac{3}{4}\right)^{j} n\right]$$

$$= cn \sum_{j} \left(\frac{3}{4}\right)^{j} E[X_{j}] \le 2cn \sum_{j} \left(\frac{3}{4}\right)^{j} \le 2cn \frac{1}{1 - \frac{3}{4}}$$

$$= 8cn = O(n)$$

### Outline

• Randomized selection algorithm

• Deterministic selection algorithm

### A Good Pivot

- Best pivot: the median
  - But, this is a circular problem
- Goal: find pivot guaranteed to be good enough
- Idea: use "median of medians"

### A Deterministic ChoosePivot

#### ChoosePivot(A, n)

- A subroutine called by the deterministic selection algorithm
- Steps:
- 1. Break A into n/5 groups of size 5 each
- 2. Sort each group (e.g., use insertion sort)
- 3. Copy n/5 medians into new array C
- 4. Recursively compute median of C
  - By calling the deterministic selection algorithm!
- 5. Return the median of C as pivot

# Deterministic Selection Algorithm

```
Dselect(int A[], int n, int i) {
      // find i-th smallest item of array A of size n
        if(n == 1) return A[1];
        Break A into groups of 5, sort each group;
        C = n/5 \text{ medians};
        p = Dselect(C, n/5, n/10);
                                          ChoosePivot
        Partition A using pivot p;
        Let j be the index of p;
Same as
        if(j == i) return p;
Rselect
        if(j > i) return Dselect(1st part of A, j-1, i);
        else return Dselect(2nd part of A, n-j, i-j);
```

The function has two recursive calls

### Short Break - 5 min

• Question Time!

#### Runtime of Dselect

- Theorem: For every input array of length n, Dselect runs in O(n) time
- Warning: not as good as Rselect in practice
  - Worse constants
  - Not-in-place: Need an additional array of n/5 medians



### What's the Runtime of Step 2?

```
Dselect(int A[], int n, int i) {
// find i-th smallest item of array A of size n
1 if(n == 1) return A[1];
2 Break A into groups of 5, sort each group;
3 C = n/5 \text{ medians};
4 p = Dselect(C, n/5, n/10);
5 Partition A using pivot p;
6 Let j be the index of p;
7 if(j == i) return p;
8 if(j > i) return Dselect(1st part of A, j-1, i);
9 else return Dselect(2nd part of A, n-j, i-j);
                     B. \Theta(n^2)
A. \Theta(n)
C. \Theta(n \log n)
                    D. \Theta(n \log \log n)
```

### Runtime of Dselect

Assume the runtime is T(n)

```
Dselect(int A[], int n, int i) {
// find i-th smallest item of array A of size n
1 if(n == 1) return A[1];
2 Break A into groups of 5, sort each group; \Theta(n)
3 C = n/5 medians; \Theta(n)
4 p = Dselect(C, n/5, n/10); T(n/5)
5 Partition A using pivot p; \Theta(n)
6 Let j be the index of p;
7 if(j == i) return p;
8 if(j > i) return Dselect(1st part of A, j-1, i);
9 else return Dselect(2nd part of A, n-j, i-j);
T(?)
```

#### Recurrence

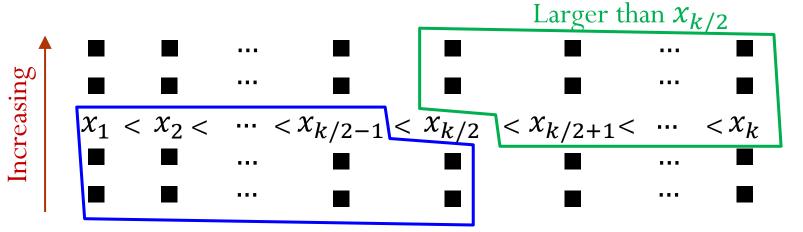
- There exists a positive constant *c* such that
  - $T(1) \leq c$
  - $T(n) \le cn + T\left(\frac{n}{5}\right) + T(?)$
- The next question is what is the size of the array of the second recursive call

### Lemma on Size

- Lemma:  $2^{\text{nd}}$  recursive call guaranteed to be on an array of size  $\leq 0.7n$  (roughly)
- (Rough) proof:
  - Let k = n/5: number of groups
  - Let  $x_i$  be the i-th smallest of the k medians
  - Thus, the pivot is  $x_{k/2}$

#### Proof of Lemma

• Imagine we layout elements of A in a 2-D grid



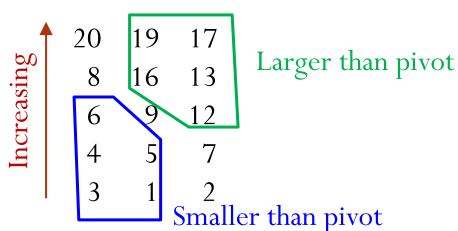
Smaller than  $x_{k/2}$ 

- At least  $\sim (3/5)*(1/2) = 30\%$  elements smaller than  $x_{k/2}$
- At least  $\sim 30\%$  elements larger than  $x_{k/2}$
- Result: Number of elements  $< x_{k/2}$  is in between 30% and 70%. The same for number of elements  $> x_{k/2}$

### Example

• Input:

After sorting each group of 5 elements



### Recurrence

- There exists a positive constant *c* such that
  - $T(1) \leq c$
  - $T(n) \le cn + T\left(\frac{n}{5}\right) + T\left(\frac{7n}{10}\right)$
- <u>Note</u>: different-sized sub-problems. Cannot use master method!
- How can we solve this?
  - <u>Strategy</u>: Hope and check
- Hope: there is a constant a (independent of n) such that  $T(n) \le an$  for all n > 1
  - Then T(n) = O(n)
- We choose a = 10c

# Proof T(n) = O(n)

- <u>Claim</u>: suppose there exists a positive constant *c* such that
  - 1.  $T(1) \le c$

2. 
$$T(n) \le cn + T\left(\frac{n}{5}\right) + T\left(\frac{7n}{10}\right)$$
  
Then  $T(n) \le 10cn$ 

- Proof by induction
  - Base case:  $T(1) \leq 10c$
  - Inductive step: inductive hypothesis  $T(k) \leq 10ck$ ,  $\forall k < n$ . Then

$$T(n) \le cn + T\left(\frac{n}{5}\right) + T\left(\frac{7n}{10}\right) \le cn + 2cn + 7cn = 10cn$$

Dselect runs in linear time