

# Randomized Algorithms

## — Randomized QuickSort & $k$ -Smallest Selection

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## Credits for the resource

- The slides are based on the textbooks:
  - *Rajeev Motwani and Prabhakar Raghavan: Randomized Algorithms. Cambridge University Press. 1995.*

# Outline

1 Randomized QuickSort

2 Randomized  $k$ -Smallest Selection

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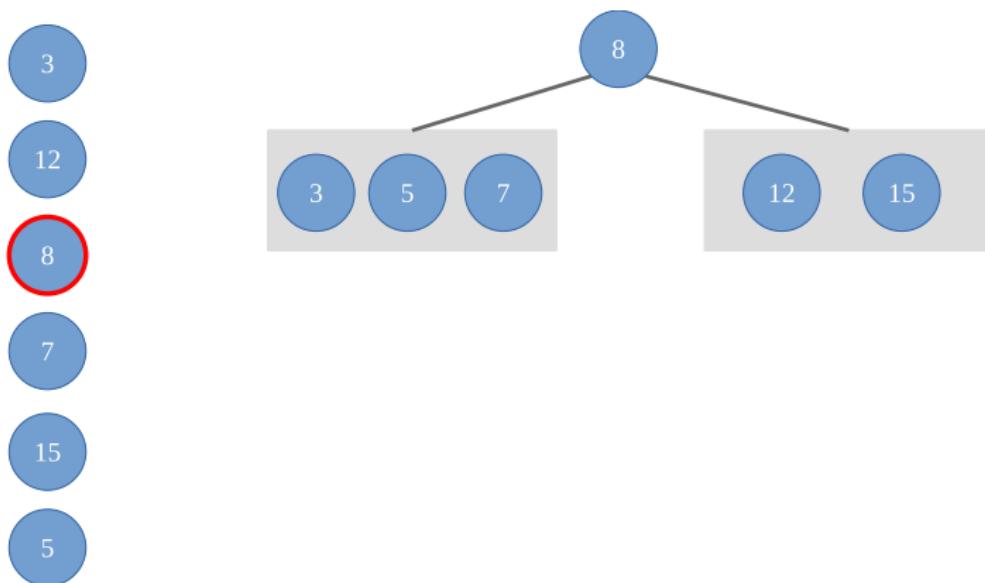
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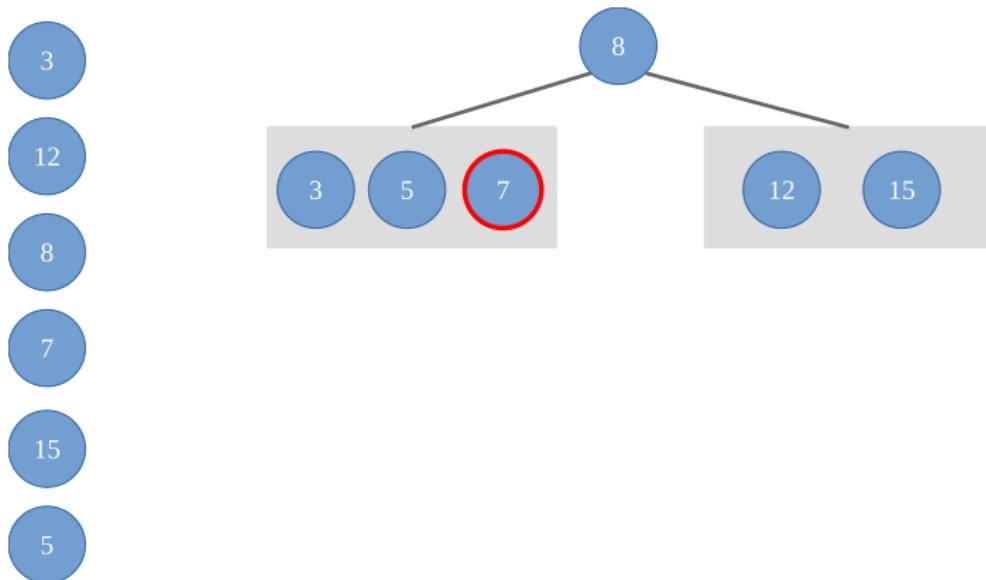
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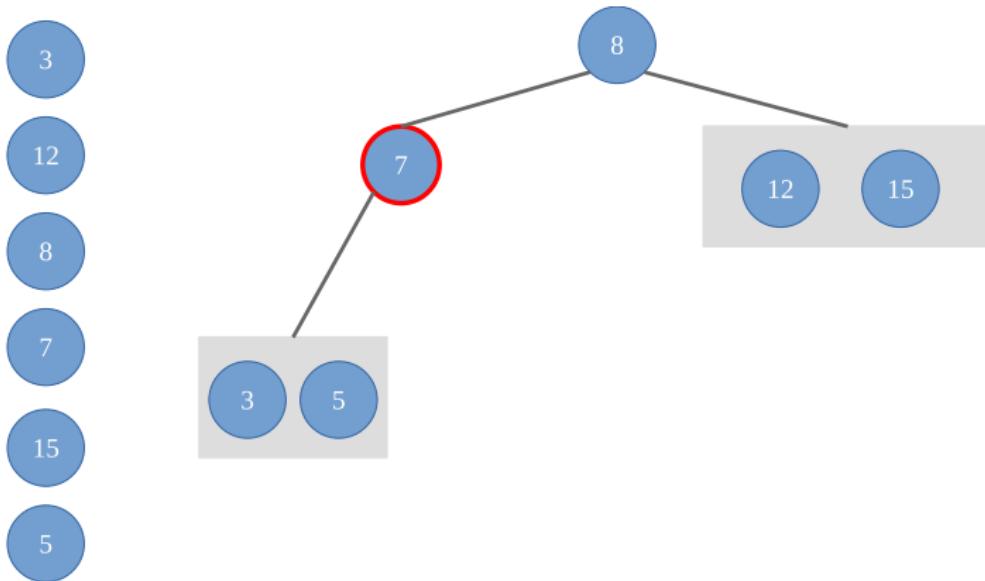
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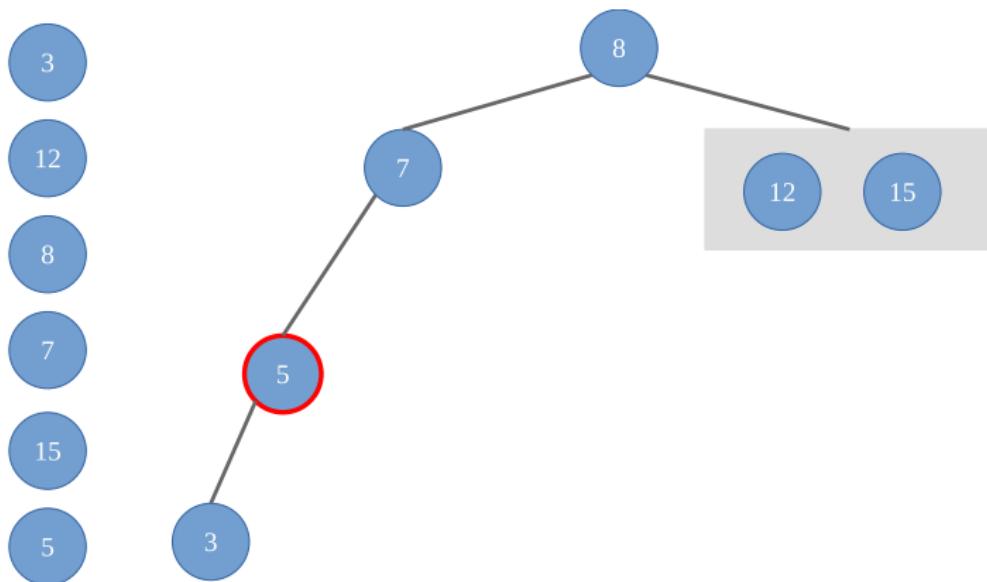
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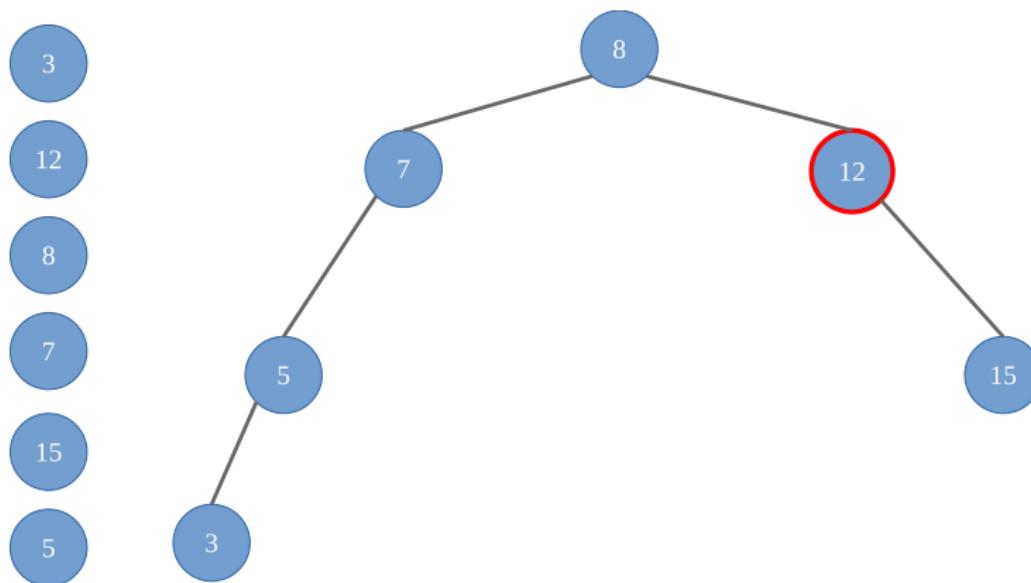
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# Algorithm RandQS

**Input:** A set of (distinct) numbers  $S$

**Output:** The elements of  $S$  sorted in increasing order.

- ① Choose an element  $y \in S$  uniformly at random;
- ② By comparing each element of  $S$  with  $y$ , compute
  - $S_1 := \{x \in S : x < y\}$ ;
  - $S_2 := \{x \in S : x > y\}$ ;
- ③ Recursively sort  $S_1$  (i.e., run  $\text{RandQS}(S_1)$ ) and  $S_2$  (i.e., run  $\text{RandQS}(S_2)$ ), and output the sorted version of  $S_1$ , followed by  $y$ , and then the sorted version of  $S_2$ .

## Analysis (Expected Number of Comparisons)

- Comparisons are performed in Step 2.
- Let  $S_{(i)}$  denote the element of rank  $i$  (i.e., the  $i$ th smallest in  $S$ ).
- Define  $X_{ij}$ :
  - $X_{ij} = 1$  if  $S_{(i)}$  and  $S_{(j)}$  are compared in an execution.
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$$\mathbb{E} \left[ \sum_{i=1}^n \sum_{j>i} X_{ij} \right] = \sum_{i=1}^n \sum_{j>i} \mathbb{E}[X_{ij}].$$

## Analysis (contd.)

- Let  $p_{ij}$  denote the probability that  $S_{(i)}$  and  $S_{(j)}$  are compared in an execution.

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- Note:  $S_{(i)}$  and  $S_{(j)}$  are compared in an execution only when one of them is an ancestor of the other in the binary tree  $T$ .

## Analysis (contd.)

$$\sum_{i=1}^n \sum_{j>i} p_{ij} = \sum_{i=1}^n \sum_{j>i} \frac{2}{j-i+1}$$

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$$\begin{aligned}\sum_{i=1}^n \sum_{j>i} p_{ij} &= \sum_{i=1}^n \sum_{j>i} \frac{2}{j-i+1} \\ &= \sum_{i=1}^n \sum_{j=i+1}^n \frac{2}{j-i+1}\end{aligned}$$

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## Analysis (contd.)

$$\begin{aligned} \sum_{i=1}^n \sum_{j>i} p_{ij} &= \sum_{i=1}^n \sum_{j>i} \frac{2}{j-i+1} \\ &= \sum_{i=1}^n \sum_{j=i+1}^n \frac{2}{j-i+1} \\ &= \sum_{i=1}^n \sum_{k=2}^{n-i+1} \frac{2}{k} \\ &\leq 2 \sum_{i=1}^n \sum_{k=1}^n \frac{1}{k} . \end{aligned}$$

## Analysis (contd.)

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 \sum_{i=1}^n \sum_{j>i} p_{ij} &= \sum_{i=1}^n \sum_{j>i} \frac{2}{j-i+1} \\
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 \end{aligned}.$$

- Note that  $H_n = \sum_{k=1}^n 1/k$

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 &= \sum_{i=1}^n \sum_{k=2}^{n-i+1} \frac{2}{k} \\
 &\leq 2 \sum_{i=1}^n \sum_{k=1}^n \frac{1}{k} = O(n \log n).
 \end{aligned}$$

- Note that  $H_n = \sum_{k=1}^n 1/k \approx \Theta(\ln n)$ .

## Exercise (3%)

### Using $O(n)$ Median-of-Medians Algorithm

- **Remark:** The Median-of-Medians algorithm (reference [here](#)) by Blum et al. can compute a median of an array of  $n$  numbers in a list in  $O(n)$  time deterministically.
- Please prove that Algorithm MedianQS (next page) can sort an array of  $n$  numbers in  $O(n \log n)$  time deterministically.

# Algorithm MedianQS

**Input:** A set of (distinct) numbers  $S$

**Output:** The elements of  $S$  sorted in increasing order.

- ① Compute the median  $y$  of  $S$  using the Median-of-Medians algorithm;
- ② By comparing each element of  $S$  with  $y$ , compute
  - $S_1 := \{x \in S : x < y\}$ ;
  - $S_2 := \{x \in S : x > y\}$ ;
- ③ Recursively sort  $S_1$  (i.e., run  $\text{MedianQS}(S_1)$ ) and  $S_2$  (i.e., run  $\text{MedianQS}(S_2)$ ), and output the sorted version of  $S_1$ , followed by  $y$ , and then the sorted version of  $S_2$ .

# Outline

1 Randomized QuickSort

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## Algorithm Rand-( $k$ )-Select

**Input:** A set of  $n$  (distinct) numbers  $S$

**Output:** The  $k$ -th smallest element of  $S$ .

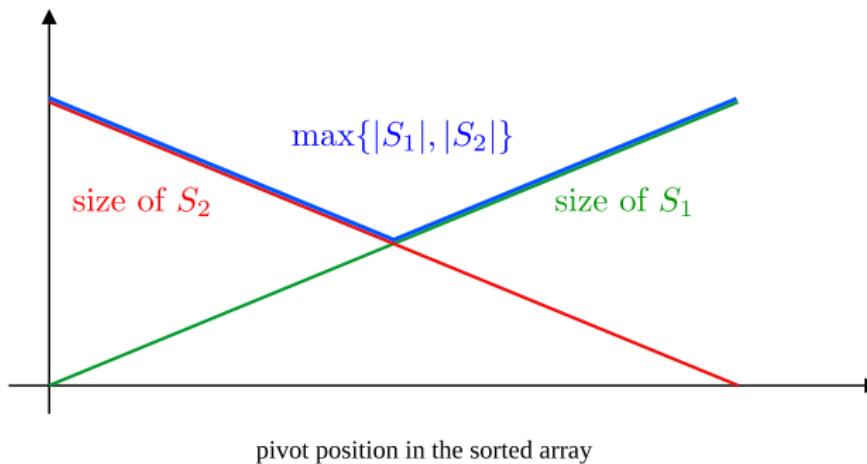
- ① Choose an element  $y \in S$  uniformly at random;
- ② By comparing each element of  $S$  with  $y$ , compute
  - $S_1 := \{x \in S : x < y\}$ ;
  - $S_2 := \{x \in S : x > y\}$ ;
- ③ If  $|S_1| = k - 1$  then return  $y$
- ④ Else
  - if  $|S_1| \geq k$ , then recursively run Rand-( $k$ )-Select( $S_1$ ).
  - else, recursively run Rand-( $k - |S_1| - 1$ )-Select( $S_2$ ).

# Time Complexity Analysis (1/3)

- Let  $X := \max\{|S_1|, |S_2|\}/n$ .

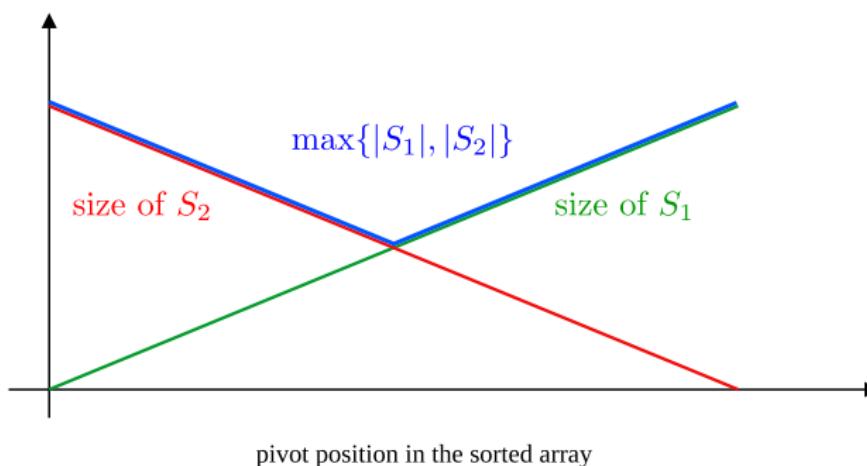
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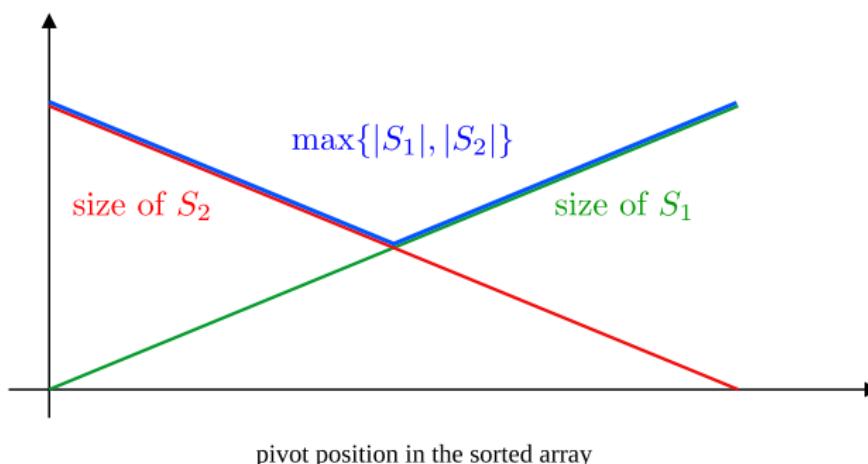
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- What's  $\mathbb{E}[X]$ ?

## Time Complexity Analysis (1/3)

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- What's  $\mathbb{E}[X]$ ?
- Prove that  $\mathbb{E}[X] \leq \frac{3}{4}$  (Exercise (1%)).

## Time Complexity Analysis (2/3)

**Note:** The recursion only runs in exactly one of  $S_1$  and  $S_2$ .

- Let  $Y_i$  be the size of the subset of  $S$  that the  $i$ th recursion proceeds with.

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- $\mathbb{E}[Y_i] =$

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- Let  $Y_i$  be the size of the subset of  $S$  that the  $i$ th recursion proceeds with.

- $$\mathbb{E}[Y_i] = \mathbb{E} \left[ n \prod_{j=1}^i X_j \right] =$$

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- $$\mathbb{E}[Y_i] = \mathbb{E} \left[ n \prod_{j=1}^i X_j \right] = n \prod_{j=1}^i \mathbb{E}[X_j] \leq n \left( \frac{3}{4} \right)^i.$$

## Time Complexity Analysis (3/3)

- Since the “partitioning” step takes  $c_1(|S|) + c_2$  for some constants  $c_1, c_2 \in \mathbb{R}$ , the expected running time of the algorithm is at most

$$\begin{aligned}\mathbb{E}[\text{Rand-}(k)\text{-Select}(S)] &\leq \sum_{i=0}^n \left( c_1 n \left(\frac{3}{4}\right)^i + c_2 \right) \\ &\leq c_1 n \left( \sum_{i=0}^n \left(\frac{3}{4}\right)^i \right) + c_2 n \\ &\leq 4c_1 n + c_2 n \\ &= O(n).\end{aligned}$$

# Discussions