

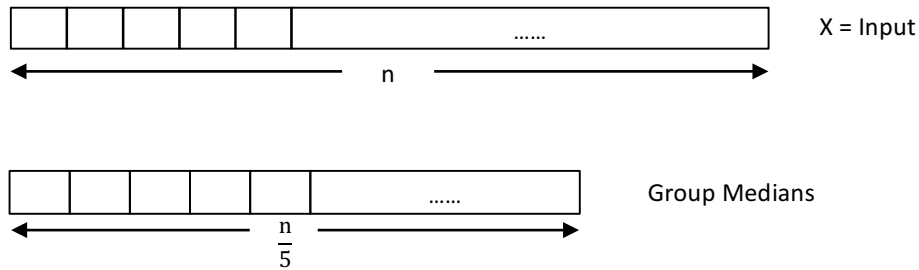
CSE 4502/5717 Big Data Analytics

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Lecture 5 – 02/05/2018

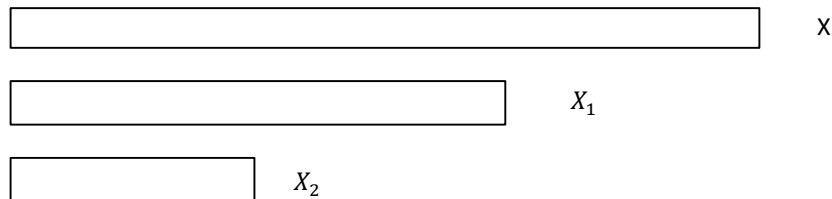
An out-of-core implementation of BFPRT algorithm:

1. Bring one block at a time and group the blocks into groups of size 5 each. Then, find the median of each group and write the medians in an output buffer.
 - When we are done with one block, bring the next block.
 - When the output buffer has one block, write it in the disk.
 - Proceed similarly until all the blocks have been processed.



The number of (read) I/O operations = $\frac{n}{B}$

2. Find recursively the median of the medians.
3. Partition the input into X_1 and X_2 , using M as the pivot.



The number of I/O operations = $\frac{n}{B}$

4. Do a recursively selection on X_1 or X_2 (as needed).

Analysis:

Let $I(n)$ be the number of I/O operations needed on any input of size n and for any i .

$$I(n) = \frac{n}{B} + I\left(\frac{n}{5}\right) + \frac{n}{B} + I\left(\frac{7}{10}n\right)$$

Hypothesis: $I(n) \leq \frac{C \cdot n}{B}$ for some constant C .

Proof by induction:

Base case: Easy

Induction step: Assume the hypothesis for inputs of size up to $(n - 1)$.

We will prove it for n :

$$I(n) \leq I\left(\frac{n}{5}\right) + I\left(\frac{7}{10}n\right) + 2 \cdot \frac{n}{B}$$

$$\leq \frac{C \cdot n}{5B} + \frac{C \cdot 7}{10} \cdot \frac{n}{B} + 2 \cdot \frac{n}{B}$$

$$\text{RHS} = 0.9 \frac{C \cdot n}{B} + 2 \cdot \frac{n}{B}$$

$$\text{RHS} \leq \frac{C \cdot n}{B} \text{ if } 0.9 \frac{C \cdot n}{B} + 2 \cdot \frac{n}{B} \leq C \cdot n$$

$$\Rightarrow 0.1C \geq 2$$

$$\Rightarrow C \geq 20$$

$$\Rightarrow I(n) \leq 20 \cdot \frac{n}{B}$$

Chernoff Bounds:

A Bernoulli trial has two outcomes: success or failure.

Assume Prob. [Success] = p

The number of successes in n independent Bernoulli trials is a Binomial Random Variable denoted as $B(n, p)$.

If $X = B(n, p)$, then:

$$(1) \text{ Prob. } [X > m] \leq \left(\frac{np}{m}\right)^m \cdot e^{-np+m}, \text{ for any } m > np;$$

- (2) Prob. $[X > (1 + \varepsilon)np] \leq \exp(\frac{-\varepsilon^2 np}{3})$, for any $0 < \varepsilon < 1$; and
 (3) Prob. $[X < (1 - \varepsilon)np] \leq \exp(\frac{-\varepsilon^2 np}{2})$, for any $0 < \varepsilon < 1$.

Example:

$$X = B(1000, \frac{1}{2})$$

$$(1 + \varepsilon) \cdot 500 = 600$$

$$\Rightarrow \varepsilon = \frac{1}{5}$$

$$\begin{aligned} \text{Prob. } [X > 600] &\leq \exp(-\frac{1}{25 \cdot 3} \cdot 500) \\ &= \exp(-\frac{20}{3}) \end{aligned}$$

$$\text{Markov's inequality: Prob. } [X > 1.2 (500)] \leq \frac{1}{1.2} = \frac{5}{6}$$

Let X be any sequence of n real numbers;

Let S be a random sample from X , with $|S| = s$;

Let $q \in S$ such that $\text{Rank}(q, S) = j$

$$\text{Note: Rank}(x, X) = |\{q \in X : q < x\}| + 1$$

Let r_j be the rank of q in X .

$$\text{Then, } E[r_j] = j \cdot \frac{n}{s}$$

Lemma (Rajasekaran & Reif 1986):

$$\text{Prob. } [|r_j - j \cdot \frac{n}{s}| > \sqrt{3\alpha} \frac{n}{\sqrt{s}} \sqrt{\log n}] \leq n^{-\alpha}$$

A Randomized Algorithm (Floyd & Rivest 1975):

1. Pick a random sample S from X .
2. Identify two elements l_1 and l_2 such that

$$\text{Rank}(l_1, S) = i \cdot \frac{s}{n} - \delta$$

$$\text{Rank}(l_2, S) = i \cdot \frac{s}{n} + \delta, \text{ where } \delta = \sqrt{4\alpha s \log n}$$

Example:

$$X = 3, 8, 12, 9, 5, 4, 11, 35, 2$$

$$\text{Rank}(5, X) = 3 + 1 = 4$$

Let $n_1 = |\{q \in X : q < l_1\}|$

Let $n_2 = |\{q \in X : q \leq l_2\}|$

If the i^{th} smallest element of X is not within $[l_1, l_2]$, then start all over. Note that the i^{th} smallest element of X will be in the interval if $i > n_1$ and $i \leq n_2$.

If the number of elements of X that are in interval $[l_1, l_2]$ is “large”, then start all over.

3. Scan through X to get $Y = \{q \in X : l_1 \leq q \leq l_2\}$.
4. Find the $(i-n_1)^{\text{th}}$ smallest element of Y and output.