

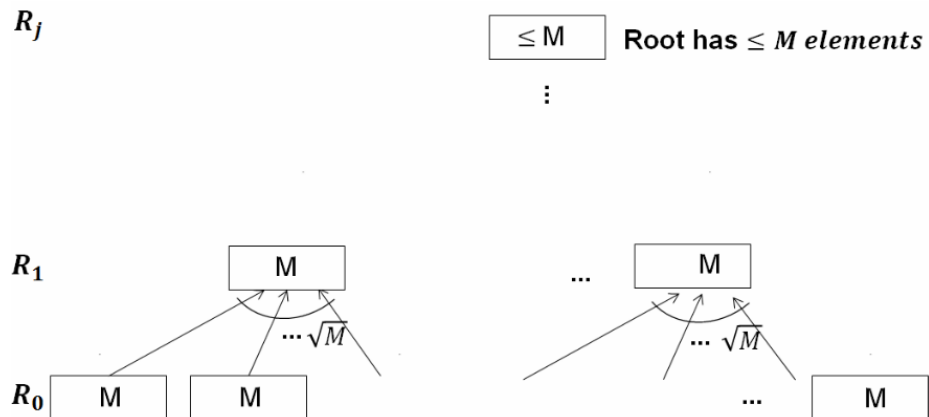
# CSE 5095: Research Topics in Big Data Analytics 2/4/2014

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## Deterministic out-of-core Selection

*input:* A sequence  $X = k_1, k_2, \dots, k_N$  and an integer  $i, 1 \leq i \leq N$   
*output:* the  $i$ -th smallest element of  $X$

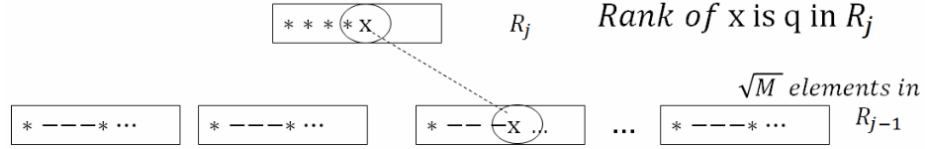
We will employ deterministic sampling. Recall that the BFPRT algorithm employs a simple form of deterministic sampling. In any out-of-core algorithm, we typically are interested in counting the number of I/O operations and we are not concerned with the computation time.



Think of a tree where the degree is  $\sqrt{M}$  and each leaf has  $M$  elements. First we sort each leaf and every leaf sends its keys with ranks  $\sqrt{M}, 2\sqrt{M}, \dots, M$  to its parent. We continue this process in every node until we reach a node that has  $\leq M$  elements. This node is the Root. The leaves are at level 0 and let the level of the root be  $j$ .

Without loss of generality lets assume that  $|R_j| = M$ .  
 In the following lines we are going to describe one level of sampling.  
 Pick  $l_1$  &  $l_2$  from  $R_j$  such that  
 $Rank(l_1, R_j) = \frac{i|R_j|}{N} - \delta$  and  $Rank(l_2, R_j) = \frac{i|R_j|}{N} + \delta$   
 $\delta$  is going to be defined in the analysis.  
 Let  $x$  be an element of  $R_j$  whose rank in  $R_j$  is  $q$ . In this case, what is  $Rank(x, R_0)$ ?

We know that  $|R_1| = \frac{N}{\sqrt{M}}$ ,  $|R_2| = \frac{N}{(\sqrt{M})^2} = \frac{N}{M}$ , ...,  $|R_j| = \frac{N}{(\sqrt{M})^j} = M \Rightarrow N = M^{\frac{j+2}{2}} \Rightarrow j = 2c - 2$ , where  $c = \frac{\log N}{\log M}$ .



$Rank(x, R_{j-1}) \geq q\sqrt{M}$   
 Also,  $Rank(x, R_{j-1}) \leq q\sqrt{M} + (\sqrt{M} - 1)^2$ .  
 This is because there is an uncertainty of  $\sqrt{M} - 1$  contributed by every node (except one) in level  $(j - 1)$  to the rank of  $x$ .

Let  $U(i)$  be the rank of  $x$  in  $R_i$ .

$$\frac{U(i+1)keys \leq x \quad \text{level } (i+1)}{U(i)keys \leq x \quad \text{level } i}$$

$U(i) \geq U(i+1)\sqrt{M}$ ,  
 $U(i) \leq U(i+1)\sqrt{M} + M^{\frac{j-i}{2}}\sqrt{M}$   
 Note that the number of nodes in level  $i$  is  $(\sqrt{M})^{j-i} = M^{\frac{j-i}{2}}$  and each such node contributes an uncertainty of  $\sqrt{M} - 1$  to the rank of  $x$ .

By repeated substitutions we have:

$$U(i) \leq qM^{\frac{j-i}{2}} + (j-i)M^{\frac{j-i+1}{2}}$$

and so:

$$U(0) \leq qM^{\frac{j}{2}} + jM^{\frac{j+1}{2}} = q\frac{N}{M} + (2c-2)\frac{N}{\sqrt{M}} \quad (1)$$

i.e.,

$$\text{Rank}(x, R_0) \in [q \frac{N}{M}, q \frac{N}{M} + (2c - 2) \frac{N}{\sqrt{M}}] \quad (2)$$

We can pick  $l_1$  such that  $\text{Rank}(l_1, R_j) = i \frac{|R_j|}{N} - (2c - 2 + \epsilon) \sqrt{M}$   
and  $l_2$  such that  $\text{Rank}(l_2, R_j) = i \frac{|R_j|}{N} + (2c - 2 - \epsilon) \sqrt{M}$ .

Using equations 1 and 2,  $\text{Rank}(l_1, R_0) \in [i - (2c - 2 + \epsilon) \frac{N}{\sqrt{M}}, i - \epsilon \frac{N}{\sqrt{M}}]$  and  $\text{Rank}(l_2, R_0) \in [i + (2c - 2 + \epsilon) \frac{N}{\sqrt{M}}, i + (4c - 4 + \epsilon) \frac{N}{\sqrt{M}}]$ .

As a result, we see that the number of input keys whose values are in the range  $[l_1, l_2]$  is  $\leq (6c - 6 + 2\epsilon) \frac{N}{\sqrt{M}}$ .

This is how one level of sampling works. The actual selection algorithm uses the above sampling process as a building block.

### Algorithm

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To begin with all the input keys are alive;  $n \leftarrow N$ ;
/*  $n$  is the number of alive keys */
repeat
  do one level of sampling;
  compute  $l_1$  and  $l_2$  as described above;
  eliminate all the alive keys that are not in the range  $[l_1, l_2]$ ;
  Adjust  $n$  and  $i$  accordingly;
  if  $n \leq M$  then
    quit the loop
  end if
until forever
perform an appropriate selection on the alive keys and output the
correct key

```

### ANALYSIS:

The number of alive keys reduces by a factor of  $\Omega(\sqrt{M}/c) = \Omega\left(\frac{\sqrt{M} \log M}{\log n}\right)$  in each iteration of the Repeat loop. If  $N$  is a polynomial in  $M$ , then the number of iterations is  $O(1)$ .

In any iteration of the repeat loop the number of I/O operations is:  
 $\frac{n}{B} + \frac{n}{B\sqrt{M}} + \frac{n}{BM} + \dots =$   
 $= \frac{n}{B} (1 + \frac{1}{\sqrt{M}} + \frac{1}{M} + \dots) \leq \frac{n}{B} (\frac{1}{1 - \frac{1}{\sqrt{M}}}) = O(\frac{n}{B})$

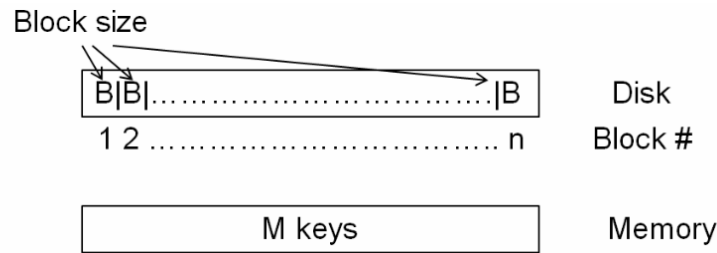
Lemma: We can do out-of-core selection deterministically in  $O(\frac{N}{B})$  I/O operations.

## A lower bound on I/O operations required to sort $N$ keys

Lemma: Sorting  $N$  keys needs  $\Omega\left(\frac{N \log \frac{N}{B}}{B \log \frac{M}{B}}\right)$  I/O operations.

Proof: Assume:

1. No new keys are generated
2. The disk is thought of as consisting of  $n = \frac{N}{B}$  blocks and the I/O's are done only with respect to these blocks



To begin with, there are  $N!$  permutations such that the correct answer could be any one of these. The number of permutations that can be generated in one I/O is  $\binom{M}{B}B!$ .

After one I/O the number of permutations remaining for the adversary is reduced to  $\frac{N!}{\binom{M}{B}B!}$ .

So we can see that the number of permutations remaining after  $t$  I/O operations is  $\frac{N!}{(\binom{M}{B}B!)^t}$ . However, when  $t > n$ , the  $B!$  term vanishes since there are only  $n$  I/O operations in which we can bring an unseen block from the disk to the core memory.

$\Rightarrow$  number of permutations remaining after  $t$  operations is  $\leq \frac{N!}{\binom{M}{B}^t (B!)^{\frac{N}{B}}}$

we want this to be  $\leq 1$ .  $\Rightarrow N! < \binom{M}{B}^t (B!)^{\frac{N}{B}}$ .

We'll use the following (crude) approximations:  $\log(x!) \approx x \log x$  and  $\log \binom{a}{b} \approx b \log \frac{a}{b}$ .

$$\begin{aligned} \text{So, } N \log N &\leq t \log \binom{M}{B} + \frac{N}{B} \log(B!) = tB \log \frac{M}{B} + \frac{N}{B} B \log B \Rightarrow \\ N \log \frac{N}{B} &\leq tB \log \frac{M}{B} \Rightarrow \\ t &\geq \frac{N}{B} \frac{\log \frac{N}{B}}{\log \frac{M}{B}} \blacksquare \end{aligned}$$