# Sorting atomic items

Chapter 5

Lower bounds

## Sorting and permuting

In RAM model Sorting includes Permuting since we need to determine the sorted permutation and then permute the items. Sorting is  $\Theta(n\log n)$  while permuting is  $\Theta(n)$ .

In disk model Sorting problem is equivalent to Permuting problem by the point of view of I/O complexity.

Moving elements is difficult as Sorting in this model. It is the real bottleneck: I/O bottelneck.

How to use Sort to Permute

# Use Sort to Permute

Permute Sequence S, S[1,n] according  $\Pi$ [1,n], i.e. Output S[ $\Pi$ (1)], S[ $\Pi$ (2)], ...S[ $\Pi$ (n)]

**RAM model**: jump on the memory to read  $S[\Pi(i)]$  then O(n). Same algorithm on 2-level model: O(n) I/O's: Too much!

Use Sort and Scan to Permute;

1) Create sequence P of pairs  $(S[i], \Pi(i))$ 

2)Sort according  $\Pi$  component

3)Parallel scan of of S and P and change S[i] with S[Π(i)]4)Sort P on the first component

## Use Sort to Permute

- S: a, b, c, d Π: 2, 4, 1, 3 RESULT: b, d, a, c
- 1. Create P.
  - P: <a, 2> <b, 4> <c, 1> <d, 3>
- 2. Sort on  $\Pi$  component
  - P: <c, 1> <a, 2> <d, 3><b, 4>
- 3 Parallel scan of S and P to substitute to  $\Pi[i]$ , S[ $\Pi(i)$ ] in P
  - S: a, b, c, d
  - P: <c, 1> <a, 2> <d, 3><b, 4>

P: <c, a> <a, b> <d, c> <b, d>

4 Sort on the first component

P: <a, b> <b, d> <c, a> <d,c>

## Use Sort to Permute

Algorithm uses 2 Scan and 2 Sort. Hence:

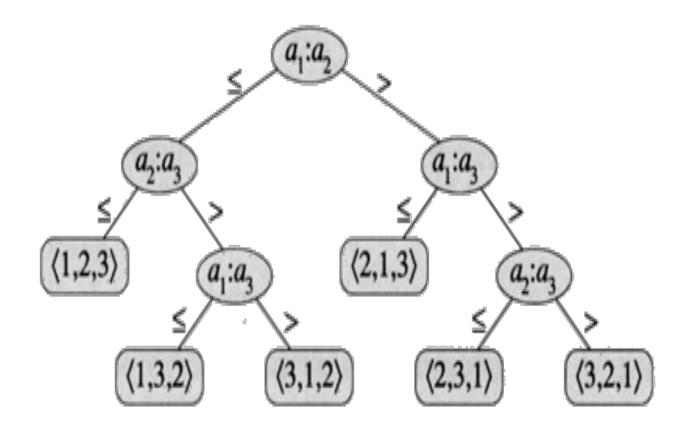
 $O(\min\{n, (n/B) \log(n/M)\})$  I/O's

This bound and that for Sorting are optimal for I/O's. The bounds are equal whenever  $n = \Omega(n/B) \log(n/M)$ 

	time complexity (RAM model)	I/O complexity (two-level memory model)
Permuting	O(n)	$O(\min\{n, \frac{n}{B} \log_{M/B} \frac{n}{M}\})$
Sorting	$O(n \log_2 n)$	$O(\frac{n}{B}\log_{\frac{M}{R}}\frac{n}{M})$

# Lower bound for sorting

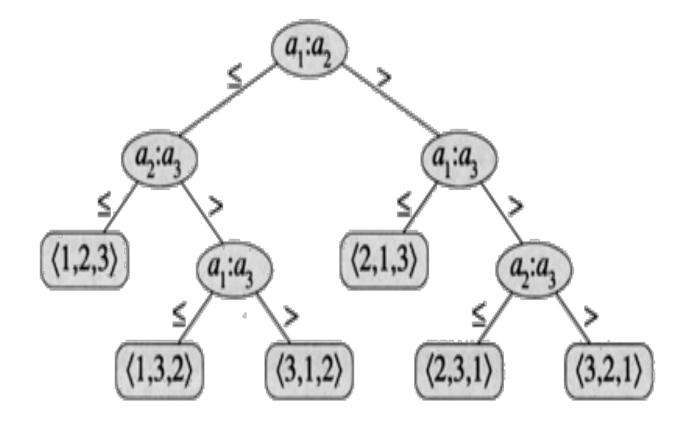
RAM model: Comparison tree to prove lower bound. Node: comparison. Leaf: solution. Root-to-leaf path t: execution of the alg. on specific data.



# Lower bound for sorting

Sorting: binary tree. The possible solutions (n! for sorting) must be allocated on the leaves.  $2^+ \ge n!$ 

 $T_2 \log(n!) = \Omega (nlogn)$ 



### Lower bound sorting in 2 level model

#### Comparison tree.

Account for I/O operations.

Operations in internal memory can be used for free.

Every node of the decision tree corresponds to one I/O. The fan-out corresponds to the result of the comparisons

after an I/O:

A block of B new elements is fetched to M. M-B elements are old , B are new. The B elements can occupy M positions of M . B

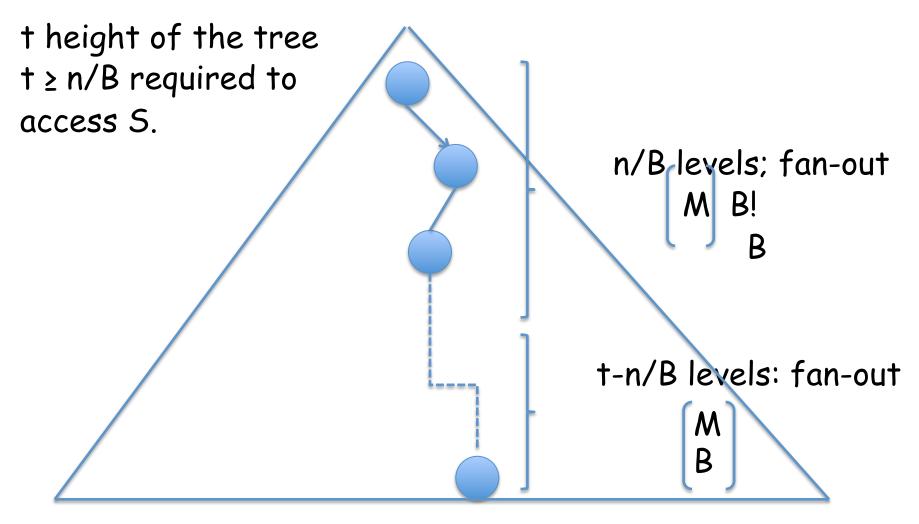
### Lower bound sorting in 2 level model

An I/O operations can generate M different results. B

Un addition, we have to consider B! different permutations of B. (the other M-B items have already been considered in previous I/O operations).

In total  $\begin{bmatrix} M \\ B \end{bmatrix}$  B! possible orderings generated by an I/O operation and by the internal comparisons.

### Lower bound sorting in 2 level model



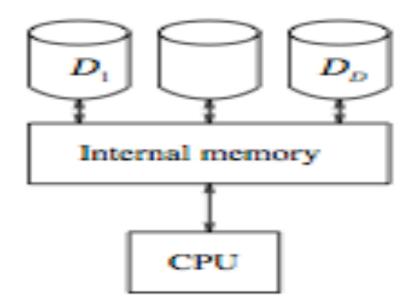
The number of leaves is  $\begin{bmatrix} M \\ B \end{bmatrix}^{\dagger} (B!)^{n/B}$ 

# Sorting: lower bound $\binom{M}{B}^t (B!)^{\frac{N}{B}}$ N! $\geq$ $t \log \binom{M}{B} + \frac{N}{B} \log(B!) \ge \log(N!)$ $t B \log(\frac{M}{B}) + \frac{N}{B} B \log B \ge N \log N$ $tB\log(\frac{M}{B})$ $\geq N \log(\frac{N}{R})$ $\geq \frac{N}{B} \frac{\log(\frac{N}{B})}{\log(\frac{M}{B})}$

 $t = \Omega (N/B) \log_{M/B} (N/B) = \Omega (N/B) \log_{M/B} (N/M)$ 

Lower bound for the D disks model

- Parallel D disks model : computer + D disks
- Input/output are from disks



Lower bound for the D disks model I/O operation: 1 block of B data is fetched to the core memory of size M from each one disk. DB data are fetched in parallel.

• Evaluate the number of parallel I/O's

The previous bound can be easily extended to D disks. A comparison-based Sorting algorithm must execute:

 $\Omega((n/DB) \log_{M/B}(n/DB))$  I/O operations

### Lower bound for the D disks model

Observation: D does not appear in the base of log. If this would be the case, it will increase the bound, so penalizing the sorting algorithm which uses D disks! MergeSort is optimal for 1 disks but it is not for D disks.

The merging should be O(n/DB) I/O's, that is at each step D pages are fetched one per disk, with an I/O. Merging is not parallel: after a comparison more than B items have to be possibly fetched from the same disk.

### Sorting in the D disks model

- Disk Striping technique: data layout on disks
- Look to the D disks as a single disk B'=DB.
- The bandwith of I/O's increases but design efficient alg . is more difficult.

 $O((n/B') \log_{M/B'}(n/M)) = O((n/DB) \log_{M/DB}(n/M))$ 

- Observe: the base of log. increases and disk striping is more and more inefficient as D increases.
- Merge is as before.
- Problem: the independency of disks is not exploited they are used as a monolithic system. Very difficult to exploit it!

### Sort in the D disks model

We must design a different algorithm. In the following:

• Greed Sort: elegant and complex new algorithm achieving a close to optimal upper bound.

### Lower bound for Permuting

1 disk model:

The previous algorithm was optimal!

D disks model:  $\Omega(\min \{n/D, (n/DB)\log_{M/B}(n/DB)\})$ 

The bounds for sorting and permuting are the same except for the case: Blog(M/B)  $\leq$  logn. This inequality holds for n =  $\Omega$  (2<sup>8</sup>) (since B and M are few k bytes and few Gigabytes respectively and log (M/B) can be neglected).

This situation id unreasonable!

In practice, Sorting = Permuting