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 $\langle i, \Pi(i) \rangle$
- 2) Sort according T component
- 3) Parallel scan of of S and P and change $\Pi(i)$ with $S[\Pi(i)]$
- 4) Sort P on the first component

Use Sort to Permute

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

P: <1, b> <2, d> <3, a> <4,c>

Use Sort to Permute

Algorithm uses 2 Scan and 2 Sort. Hence:

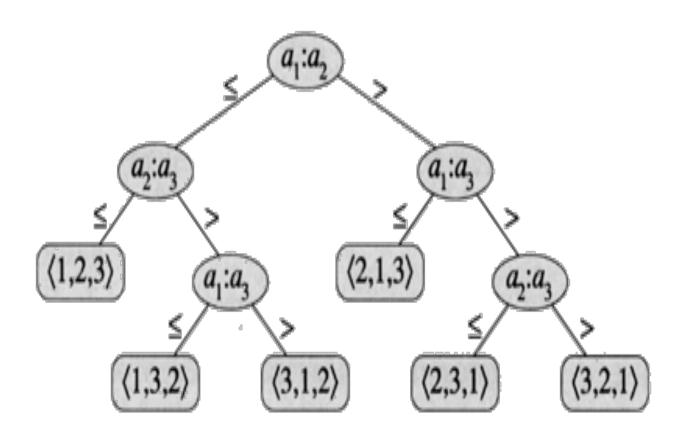
 $O(\min\{n, (n/B) \log(n/M)\})$ I/O's

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

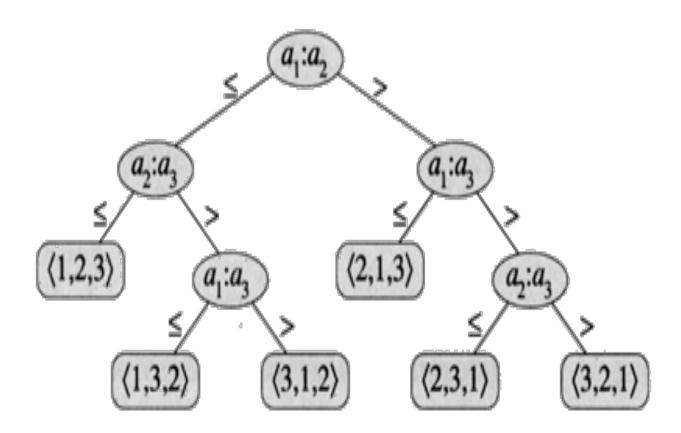
RAM model: Comparison tree to prove lower bound. Node: comparison. Leaf: solution. Root-to-leaf path t: execution of the alg. on specific data.



Lower bound for sorting

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

$$T \ge \log(n!)$$
 $t = \Omega (n \log n)$



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 B occupy positions of M in possible ways.

M B

Lower bound sorting in 2 level model

An I/O operations can generate M different results.

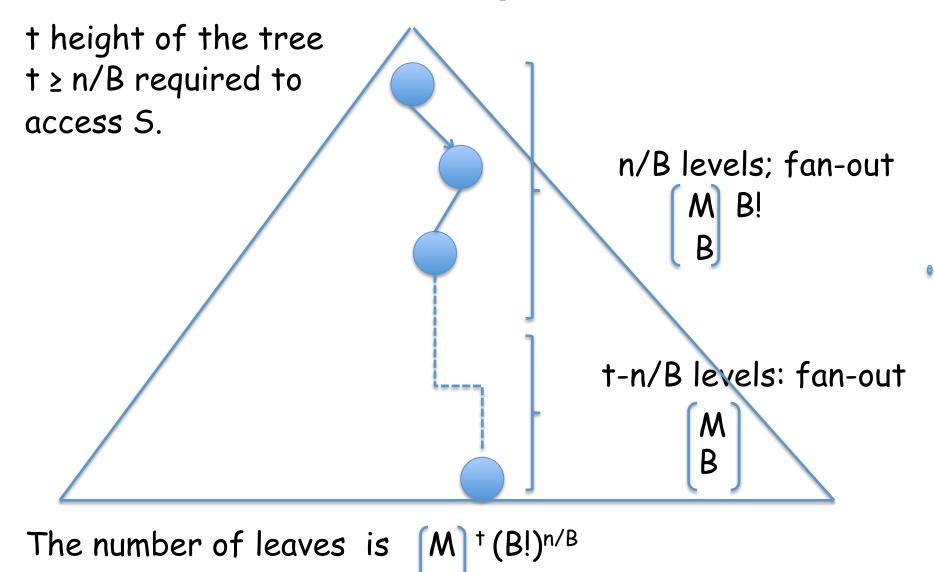
Un addition, we have to consider B! different permutations of B. (the other M-B items have already been considered in previous I/O operations).

In total [M] B! possible orderings generated by an I/O

operation and by the internal comparisons.

M B! is the fan-out of each node.

Lower bound sorting in 2 level model

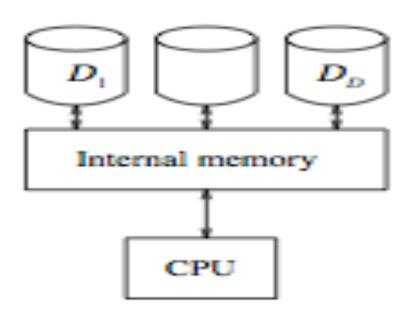


Sorting: lower bound

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

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 operation: 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.

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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:

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If B log (M/B) \leq logn then \Omega (n) otherwise \Omega (n/B)log<sub>M/R</sub>(n/M) NO PROOF
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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: $Blog(M/B) \le logn$.

This inequality holds for $n = \Omega(2^B)$

(since B and M are few k bytes and few Gigabytes respectively and log (M/B) can be neglected).

This situation is unreasonable!

In practice, Sorting = Permuting