Algorithms for Memory Hierarchies Lecture 12

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1 List ranking in $O(sort_p(N))$

Last time we discussed List ranking in $O(sort_p(N)log^*(N))$ I/O's, today we will get rid of the $log^*(N)$ factor. As a general reminder the algorithm we saw last time did the following:

- 1. Find independent set of size $\Theta(N)$
- 2. Bridge out the nodes, which are in the independent set
- 3. Solve the problem recursively on the remaining items
- 4. Bridge the nodes back in

We showed how we can do this in the PEM model using a technique called deterministic coin tossing. But still we remain with a non constant factor of $\log^*(N)$ for the I/O's.

To understand the first step of our new algorithm we look back at the list ranking algorithm from last time. We can see that deterministic coin tossing is nothing else, than a colouring of the nodes and the number of repetitions of the colouring determine the number of colors. In fact if we repeat the process t times we get an $O(\log^t(N))$ colouring. We will call this a k colouring Now we will use the induced colors and sort the nodes by color. We get groups and want the first group G_0 to contain the most nodes. Now for each group G_i we add the group to the independent set and remove all their neighbours in other groups. Trivially we can do this at start with G_0 , then with G_1 since all neighbours G_0 and G_1 have are already deleted and so on. Thus we get a valid independent set. The I/O complexity here is in $O(sort_p(N) + ksort_p(N)) = O((k+1)sort_p(N)) = O(sort_p(N)\log^t(N))$. Now it looks like we did not win anything, still there is a logarithmic factor in the I/O complexity.

1.1 Delayed Pointer processing

To get pure sorting complexity we introduce a new technique called "Delayed pointer processing". We start as above.

- 1. Run the deterministic coin toss algorithm t times. $O(tsort_p(N))$
- 2. Store with each item their colors and ID's of neighbours $O(sort_p(N))$
- 3. Group the nodes by colors, such that G_0 contains the most nodes $O(sort_p(N))$

So far nothing special happened. As above we add G_0 to the independent set and, if a node in G_0 has a neighbour in another group, add to that group a duplicate. Now for each group do the following:

- 1. Sort the group and remove the duplicated nodes $O(sort_{p(N_i)})$
- 2. Add the remaining nodes to the independent set $O(scan_p(N_i))$
- 3. Add duplicates to the appropriate groups $O(scan_p(N_i)) + \log^t(N)$

The I/O complexity can be written as:

$$O(sort_p)(N) + \sum_{i=1}^{k-1} sort_p(N_i) + \log^t(N)$$

= $sort_p(N) + \sum_{i=1}^k \frac{N}{PB} \log_{\frac{M}{B}} \frac{N_i}{B}$
 $\leq sort_p(N) + \frac{1}{PB} \log_{\frac{M}{B}} \frac{N}{B} \sum_{i=1}^k N_i + (\log^t N)^2)$
= $O(sort_p)(N)) + (\log^t N)^2)$

Now we still have a logarithmic factor in our equation, but if we choose the number of processers appropriatly we can eliminate it. The question is for which values of p is the logarithmic factor dominating. Obviously we have to solve this inequation:

$$\frac{N}{PB}\log_{\frac{M}{B}}\frac{N}{B} < (\log^t N)^2$$

Shifting it to p yields:

$$p < \frac{N}{B(\log^t(N))^2} = O(\frac{N}{B\log N}$$

2 Parallel Distribution Sweeping

The aim here is to adapt the distribution sweeping technique to multicores architectures. The I/O complexity reached on this kind of architecture with p cores is at best:

$$O(\frac{N}{B}\log_{\frac{M}{B}}(\frac{N}{B}) + \frac{k}{pB})$$

But in this lesson is going to limit to a solution whose I/O complexity is:

$$O(\frac{N+k}{B}\log_{\frac{M}{B}}(\frac{N+k}{B}) = O(sort_p(N+k))$$

Let

$$d = \min\{\sqrt{\frac{N}{p}}; \frac{M}{B}; p\}$$

2.1 General Algorithm

Algorithm:

- 1. Partition the space into d vertical slabs & p horizontal slabs with equal number of objects in each slab
- 2. Preprocess objects in each vertical slab using all processors (specific for a problem)
- 3. Sweep each horizontal slab using one processor. Process all horizontal objects that span ¿ 1 vertical slabs, distribute objects into appropriate slab lists
- 4. Recursively solve the problem on each vertical slab (allocate proportional number of processors)

Base case: one processor per slab runs sequential I/O efficient solution.

2.2 The Batched Stabbing Query Problem

Algorithm:

- 1. Partition the space into d vertical slabs & p horizontal slabs with equal number of objects in each slab
- 2. Count for each portion of horizontal segment h_i that spans σ_j how many vertical segments h_i intersects σ_j . Readjust horizontal slab boundaries such that each slab cointains $\theta(\frac{N}{p})$ vertical segment & $\theta(\frac{k}{p})$ copies of horizontal segment (using prefix sum).
- 3. Create a copy of each h_i that spans σ_j & intersect ≥ 1 vertical segment or has endpoint in σ_j in σ_j 's slab list. Create copies of vertical segments in σ_j in σ_j 's slab list
- 4. Recursively solve the problem on each vertical slab (allocate proportional number of processors)



Figure 1: Divisions in slabs

Base case: run sequential I/O efficient line/segment intersection solution.

Now we need to detail a little more the step 2:

- a Set weights in each slab σ_j as follows: +1 on bottoms end point and -1 on top end point
- b Run prefix sums: this allows to know how many an horizontal line intersects vertical lines as only sums which differs from 0 are those which are intersected:



Figure 2: Prefix sum uses to count intresections

2.3 I/O complexity per recursive call in the case of The Batched Stabbing Query Problem

- Second step: Q(N,P)
- Third step: $scan(\frac{N}{p}) + scan(\frac{k_k}{p})) \leq O(\frac{N}{pB} + \frac{k}{pB})$

I/O complexity of base case: $O(\frac{N'}{B}log_{\frac{M}{B}}(\frac{N'}{B}) + \frac{k}{B})$ With ih this case: $N' = \theta(\frac{N+k}{p})$

Total I/O complexity:

$$Total = \sum_{k=1}^{\log_{d}p} O(Q(N;p) + \frac{N+k}{pB}) + O(\frac{N+k}{pB}\log_{\frac{M}{B}}(\frac{N+k}{pB}) + \frac{k}{B})$$

$$Q(N;p) = O(\frac{N''}{pB} + logp) = O(\frac{N''}{pB}) \leqslant O(\frac{N+k}{pB})$$

If we consider that the copies have been done.

$$\begin{aligned} Total &= \sum_{k=1}^{\log_d p} O(\frac{N+k}{pB}) + O(\frac{N+k}{pB} \log_{\frac{M}{B}}(\frac{N+k}{pB}) + \frac{k}{pB}) \\ Total &= \sum_{k=1}^{\log_d p} O(\frac{N+k}{pB}) + O(\frac{N+k}{pB} \log_{\frac{M}{B}}(\frac{N+k}{pB})) \\ Total &= O(\frac{N+k}{pB} \log_d p) + O(\frac{N+k}{pB} \log_{\frac{M}{B}}(\frac{N+k}{pB})) \\ Total &= O(\frac{N+k}{pB} \log_{\frac{M}{B}}(\frac{N+k}{pB})) \\ Total &= O(sort_p(N+k)) \end{aligned}$$

In sequential I/O solution:

$$O(\frac{N}{B}\log_{\frac{M}{B}}\frac{N}{B} + \frac{k}{B}) = sort(N) + scan(k)$$