Algorithm Theory, Winter Term 2015/16 Problem Set 14 - Sample Solution

Exercise 1: All-prefix-sums on Multi-dimensional Matrices (6 points)

In this exercise, we consider a generalization of the the all-prefix-sums problem discussed in the lecture. We study the following all-prefix-sums problem defined on a d-dimensional array.

We are given a $n \times n \times \cdots \times n$ array \mathcal{A} with entries a_{i_1,i_2,\ldots,i_d} (for $i_j \in \{1,\ldots,n\}$). The goal is to calculate the all-prefix-sums array \mathcal{S} with entries s_{i_1,i_2,\ldots,i_d} which are defined as follows

$$s_{i_1,i_2,\dots,i_d} := \sum_{j_1=1}^{i_1} \sum_{j_2=1}^{i_2} \cdots \sum_{j_d=1}^{i_d} a_{j_1,j_2,\dots,j_d}$$

Note that for d = 1, the problem is the usual all-prefix-sums problem from the lecture.

- (a) (3 points) To warm up, we first consider the case d = 2. Give an efficient algorithm to solve the 2-dimensional all-prefix-sums problem. What are the work T₁ and span T∞ of your solution. *Hint:* The problem can be solved by performing 2n standard all-prefix-sums computations.
- (b) (2 points) Generalize the above algorithm to $d \ge 2$ dimensions. What are work T_1 and span T_{∞} of the resulting parallel algorithm?
- (c) (1 points) What is the minimum number of processors needed such that asymptotically, the maximum possible speed-up can be achieved?

Solution

(a) For d = 2, we can think of \mathcal{A} as a 2-dimensional matrix of order $n \times n$. The $(i, j)^{th}$ entry of \mathcal{A} is $a_{i,j}$. By the definition, the $(i, j)^{th}$ entry of all-prefix-sums matrix \mathcal{S} is $s_{i,j} := \sum_{k=1}^{i} \sum_{l=1}^{j} a_{k,l}$. That is, $s_{i,j}$ is the sum of all entries of \mathcal{A} with indices in $[1, i] \times [1, j]$. We can write $s_{i,j}$ as

$$s_{i,j} = (a_{1,1} + a_{2,1} + \dots + a_{i,1}) + (a_{1,2} + a_{2,2} + \dots + a_{i,2}) + \dots + (a_{1,j} + a_{2,j} + \dots + a_{i,j})$$

= $\bar{s}_{i,1} + \bar{s}_{i,2} + \dots + \bar{s}_{i,j}$,

where $\bar{s}_{i,k} = (a_{1,k} + a_{2,k} + \dots + a_{i,k})$ is the sum of the first *i* entries of the *k*-th column of \mathcal{A} . Therefore, $s_{i,j}$ is the sum of the first *j* entries of the *i*th row of the matrix $\bar{\mathcal{S}} = (\bar{s}_{i,j})$. Notice that any column *k* of the matrix $\bar{\mathcal{S}}$ is the all-prefix-sums of the entries in the *k*th column of \mathcal{A} . Hence, we first construct $\bar{\mathcal{S}}$ by computing the all-prefix-sums of all the columns of \mathcal{A} separately. We can do this in parallel for each column of \mathcal{A} . Since \mathcal{A} has *n* columns, we have *n* instances of all-prefix-sums problem. Using the parallel algorithm of the lecture, we can compute all the *n* all-prefix-sums in parallel.

Now the final matrix S is the all-prefix-sums of each rows of the matrix \overline{S} . Again we can use the parallel all-prefix-sums algorithm of the lecture for each rows of \overline{S} in parallel and compute S. We know that (from the lecture), for one all-prefix-sums instance, total work is $T_1 = n$ and span

is $T_{\infty} = \log n$. Here we compute 2n such instances: n for computing \bar{S} and n for computing the final S. Hence, the total work will be $T_1 = 2n^2$ and the span will be $T_{\infty} = 2\log n$, since the final S computation depends on the computation of the intermediate matrix \bar{S} (each dependant tree height is $\log n$).

- (b) The generalization is straightforward. Let us first look at the case for d = 3. We can think of it as a combination of two phases: first compute n instances of 2-dimensional all-prefix-sums problems and then compute the final all-prefix-sums array corresponding to the third dimension. We first solve these n 2-dimensional all-prefix-sums in the similar way as done in question (a). In fact, we can compute all the n instances in parallel. The work for this phase would be $n \cdot 2n^2 = 2n^3$. Then we have to compute the all-prefix-sums corresponding to the third dimension. There are again n^2 instances of 1-dimensional all-prefix-sums problems. This will give us the final allprefix-sums array for d = 3. The total work will be $T_1 = 2n^3 + n^2 \cdot n = 3n^3$. The span will be $T_{\infty} = 2\log n + \log n = 3\log n$, since the computation of the all-prefix-sums corresponding to the third dimension depends on the previous n instances of 2-dimensional all-prefix-sums computations. In the similar way, we can extend the solution for general $d \ge 2$. We can do step by step as above: first compute n instances of (d-1)-dimensional all-prefix-sums problems and then compute the final all-prefix-sums array corresponding to the d^{th} dimension. The total work will be $T_1 = dn^d$ and the span will be $T_{\infty} = d\log n$.
- (c) We use the Brent's Theorem (see the lecture). Recall that if p is the number of processors, then $p = O(\frac{T_1}{T_{\infty}})$. Therefore by putting the above values of T_1 and T_{∞} , we get $p = O(\frac{n^d}{\log n})$.

Exercise 2: Merging Two Sorted Arrays (6 points)

You are given two sorted arrays $A = [a_1, \ldots, a_n]$ and $B = [b_1, \ldots, b_n]$, each of size n. The goal is to merge them into one sorted array $C = [c_1, \ldots, c_{2n}]$ of length 2n.

- (a) (1.5 points) We first consider the following subproblem. Given an index $i \in \{1, ..., n\}$, we want to find the final position $j \in \{1, ..., 2n\}$ of the value a_i in the array C. Give a fast sequential algorithm to compute j. What is the (sequential) running time of your algorithm?
- (b) (1.5 points) Use the above algorithm to construct a parallel merging algorithm. The work T_1 of your algorithm should be at most $O(n \log n)$ and the span T_{∞} should be (asymptotically) as small as possible. What is the span T_{∞} of your algorithm?
- (c) (3 points) We now want to solve the merging problem in constant time (in parallel). Show that by using O(n) processes, the subproblem considered in (a) can be solved in O(1) time. Use this to get a constant-time parallel algorithm to merge the two sorted arrays. How many processors do you need to achieve a constant-time algorithm?

Solution

Assume that all arrays are sorted in ascending order.

(a) Note that the first (i-1) values in A are before a_i in C, because the array A is sorted. Now we have to find out how many values in B are before a_i . That is to find the largest index k such that $b_k \leq a_i$. One easy way for this is to compare a_i with the elements of B one by one starting from b_1 and find the index k. This will take O(n) time in general, since the size of the array B is n. However, we can do it faster using the divide and conquer approach (this is exactly the binary search). We recursively break the array B into two parts of equal size and check in which side a_i falls (and ignore the other side). Using divide and conquer approach we can find the index k in $O(\log n)$ time. Once we find the k, then the final position of a_i would be (i-1) + k + 1 (assuming the array indices starting from 1).

- (b) In the above algorithm, we see that one processor can find the final position of a value a_i in $O(\log n)$ time. Now we consider n processors corresponding to each value a_i in A and compute their positions in the output array C in parallel. All the processors can find the final position of every values of A in $O(\log n)$ time. Then in the same way, we compute the position of all the values of B in C using n processors and $O(\log n)$ time. Hence, we can merge the two sorted arrays into one sorted array in $O(\log n)$ time, using n processors. The total work is $T_1 = O(n \log n)$ and the span is $T_{\infty} = O(\log n)$.
- (c) Consider a particular value a_i of A and we want to find the final position of a_i in C. Let us take n processors $p_k : k = 1, 2, ..., n$. Each processor p_k compares the value a_i with two consecutive values b_{k-1} and b_k in B. All the processors do it in parallel. (Note that the array indices starting from 1, so we assume $b_0 = -\infty$ for consistency). Since the values b_k are in ascending order (sorted), there will be only one processor p_t which see that $b_{t-1} \leq a_i$ and $b_t > a_i$. That is there are exactly t-1 values in the array B which are smaller than a_i . Hence, the processor p_t can decide the final position of a_i which would be (i-1) + (t-1) + 1 (since there are i-1 values smaller than a_i in A). The processor p_t can write the value a_i safely in the final array C. Note that the processor p_n may observe that $b_n \leq a_i$, then the final position of a_i would be (i-1) + n + 1. Since all the processors computing this in parallel, it takes constant time. Also we used n processors for this. We can extend this algorithm for all the values in A using n^2 processors and they can write all the values a_i in the correct place in C. Notice that there would not be any conflicts when writing in C, since a processor p_t only writes the value in one cell of the array C. Thus we can put all the values of A in the output array C in constant time.

Now we want to put all the values of B in C. Again we can use the same approach as above i.e., we find the right index of a particular value b_j in B by comparing with values in A. We have to be a bit careful in this case. During the comparison of a value b_j with two consecutive values a_{k-1} and a_k , each processor p_k checks if $a_{k-1} < b_j$ and $a_k \ge b_j$, i.e., processors find index the t, for which $a_{t-1} < b_j$ and $a_t \ge b_j$ holds. This "strict" less inequality is necessary to avoid any concurrent writing or conflicts in C. The processor p_t which found the index t, can decide the final position of a_i as (j-1) + (t-1) + 1.

Therefore, we can merge the two sorted array of size n into one sorted array in O(1) time using n^2 processors.