### Lecture 20

# Massively Parallel Algorithms: Sorting, Counting Distinct Elements

Course: Algorithms for Big Data

Instructor: Hossein Jowhari

Department of Computer Science and Statistics Faculty of Mathematics K. N. Toosi University of Technology

Spring 2021

## MPC: recap



- N input data size, M machines
- S = o(N) memory size per machine
- $\blacktriangleright$  Each machine communicates at most S words in each round

# Sum of N integers

Each machines computes its local sum and sends it up to the coordinator.



•  $\sqrt{N} < S \le N$ 

• 
$$M = \frac{N}{S}$$

• # rounds = 
$$1$$



• 
$$2 \le S \le \sqrt{N}$$
  
•  $M = \frac{N}{S}$   
•  $\#$  rounds =  $\log_S N$ 

## Sorting N integers

Input Data: N integers  $\{a_1, a_2, \ldots, a_N\}$ 

Assumption: The input integers are distinct (no repetitions).

The input is partitioned among the machines. Each machine gets  $S = O(N^{2/3})$  elements.  $M = O(N^{1/3})$ 

Output: The rank of each element is known by some machine.



A parallel algorithm inspired by quicksort

Stage 1: selecting potential pivots

- The machines select a random subset X of the elements. Each element is picked with probability  $p = \frac{N^{1/3} \log N}{N}$ . We call these elements potentially pivot elements.
- The machines communicate their selected elements. At the end of this stage, the machines all know the random subset X.



#### Stage 2: local ranks for the potential pivots

- For each x ∈ X, the machines compute local-rank(x): how many elements in their input is smaller than x.
- For each x ∈ X, the machines send (x, local-rank(x)) to the coordinator.



#### Stage 3: computing the pivots

- ► Having received the local ranks, the coordinator computes the (global) rank of each potential pivot x ∈ X.
- For each N<sup>2/3</sup> length interval in {1,...,N}, the coordinator selects an element from the potential pivots X that has a rank with that interval.
- The coordinators sends the selected pivots to all machines.



Lemma: With high probability, there exists a potential pivot from each  $N^{2/3}$  length interval.

Proof: Each element is picked probability  $p = \frac{N^{1/3} \log N}{N}$ . In expectation, number of elements selected from a rank interval is  $\log N$ .

Using Chernoff bound, with high probability there is at least one elements from each interval.  $\hfill\square$ 

Let  $p_1, p_2, \ldots, p_{N^{1/3}}$  be the selected pivots.

Consider the real intervals:

$$I_1 = (-\infty, p_1], I_2 = (p_1, p_2], \dots, I_{N^{1/3}+1} = (p_{N^{1/3}}, +\infty)$$

Machine  $M_i$  will be responsible for the elements in the interval  ${\cal I}_i$ 

Stage 4: sending the elements to the responsible machines

Each machine, for each element y in its local memory, sends y to the machine M<sub>i</sub> where

$$y \in I_j = (p_{j-1}, p_j)$$

Each machine locally sorts the received elements.

Let  $S_i$  be the sorted lists owned by machine  $M_i$ . The final list  $S_1, S_2, \ldots, S_M$  is sorted in the increasing order.



### Round complexity of each stage

Stage 1: each machine selects O(log N) potential pivots in expectation. With high probability (Chernoff bound), number of selected pivots is O(log N).

The potential pivots are broadcasted to other machines. There are  $O(N^{1/3})$  machines. This can be done in one round. Recall the communication limit is  $S = O(N^{2/3})$ .

- Stage 2: Each machine sends O(N<sup>1/3</sup> log N) words to the coordinator. This can be done in 2 rounds using a broadcast tree of depth 2.
- Stage 3: The coordinator sends N<sup>1/3</sup> number of pivots to all machines. This can be done in 1 round.

Stage 4: The machines sends the elements to their responsible machines. Each machine has O(N<sup>2/3</sup>) elements. Each real interval has at most O(N<sup>2/3</sup>) elements in it. This stage can also be done in O(1) round.

[Theorem] There is a O(1) round MPC algorithm for sorting N numbers where each machine has  $O(N^{2/3})$  space.

Question: What about the general case when  $S = n^{\epsilon}$ ?

[Theorem] There is a  $O(1/\epsilon)$  round MPC algorithm for sorting N numbers where each machine has  $O(N^{\epsilon})$  space.

See Parallel Algorithms (Chapter 6) by Mohsen Ghaffari.

#### Question: What if the numbers are not distinct?

One idea is to perturb the numbers so that all numbers become distinct. We can do this by adding a small random  $\epsilon$  to all numbers. (With high probability all numbers are distinct now.) We sort the perturbed numbers and then scrap the added noise.

Question: A MPC algorithm for counting distinct elements?