Parallel Algorithms

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PARALLEL SORTING
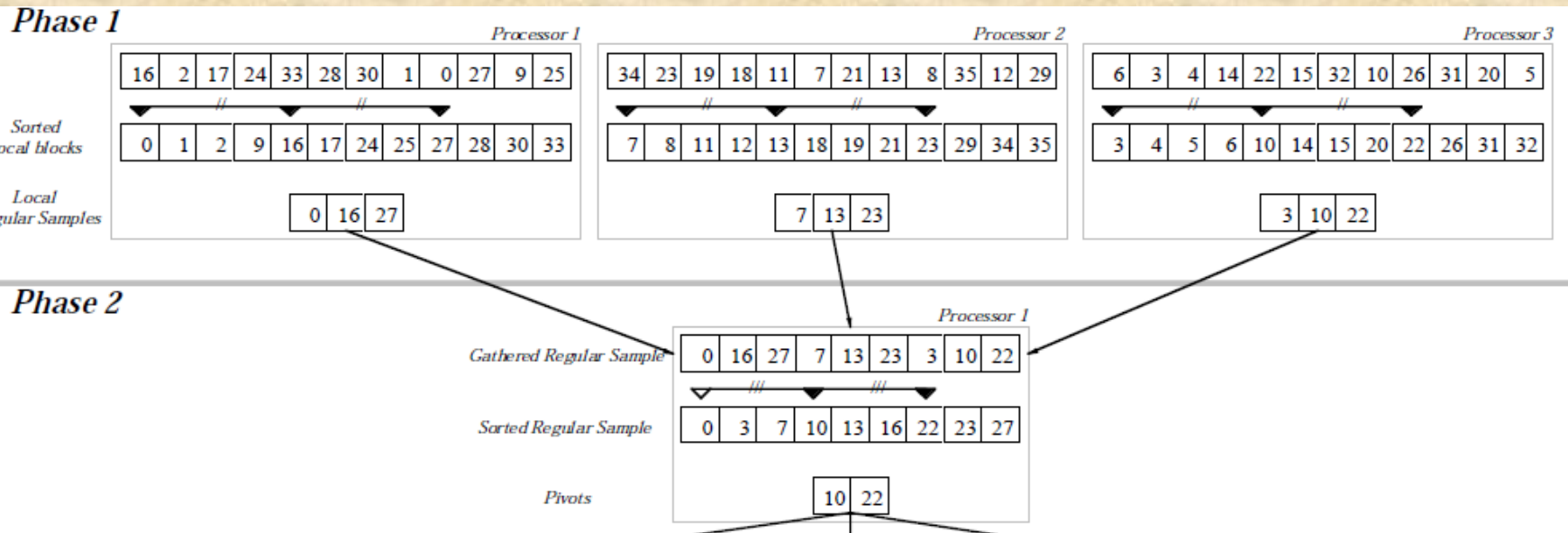
Introduction

- ☐ The input sequence of size N is distributed across P processors

- ☐ The output is such that elements in P_i is greater than elements in P_{i-1} and lesser than elements in P_{i+1}

Parallel Sorting by Regular Sampling (PSRS)

1. Each processor sorts its local data
2. Each processor selects a sample vector of size p-1; kth element is (n/p * (k+1)/p)
3. Samples are sent and merge-sorted on processor 0
4. Processor 0 defines a vector of p-1 *splitters* starting from p/2 element; i.e., kth element is p(k+1/2); broadcasts to the other processors

Example

PSRS

5. Each processor sends local data to correct destination processors based on splitters; all-to-all exchange
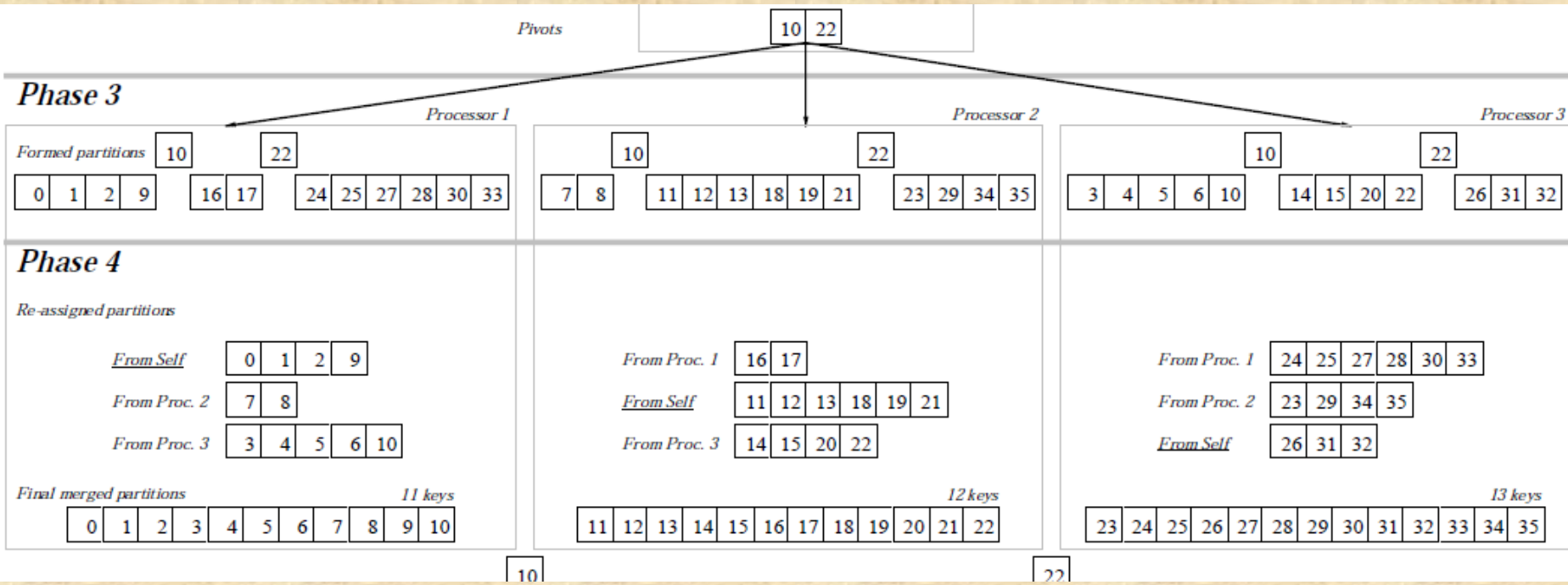6. Each processor merges the data chunk it receives

Step 5

- ☐ Each processor finds where each of the p-1 pivots divides its list, using a binary search

- ☐ i.e., finds the index of the largest element number larger than the jth pivot

- ☐ At this point, each processor has p sorted sublists with the property that each element in sublist i is greater than each element in sublist i-1 in any processor

Step 6

☐ Each processor i performs a p-way merge-sort to merge the ith sublists of p processors

Example Continued

Analysis

- ☐ The first phase of local sorting takes $O((n/p)\log(n/p))$
- ☐ 2nd phase:
  - ■ Sorting $p(p-1)$ elements in processor 0 – $O(p^2 \log p^2)$
  - ■ Each processor performs $p-1$ binary searches of n/p elements – $p\log(n/p)$
- ☐ 3rd phase: Each processor merges $(p-1)$ sublists
  - ■ Size of data merged by any processor is no more than $2n/p$ (proof)
  - ■ Complexity of this merge sort $2(n/p)\log p$
- ☐ Summing up: $O((n/p)\log n)$

Analysis

- ☐ 1st phase – no communication
- ☐ 2nd phase – p(p-1) data collected; p-1 data broadcast
- ☐ 3rd phase: Each processor sends (p-1) sublists to other p-1 processors; processors work on the sublists independently

☐ Graph Algorithms

Graph Traversal

- ☐ Graph search plays an important role in analyzing large data sets

- ☐ Relationship between data objects represented in the form of graphs

- ☐ Breadth first search used in finding shortest path or sets of paths

Parallel BFS
Level-synchronized algorithm

☐ Proceeds level-by-level starting with the source vertex

☐ Level of a vertex – its graph distance from the source

☐ Also, called **frontier-based** algorithm

☐ The parallel processes process a level, synchronize at the end of the level, before moving to the next level – Bulk Synchronous Parallelism (**BSP**) model

☐ How to decompose the graph (vertices, edges and adjacency matrix) among processors?

Distributed BFS with 1D Partitioning

☐ Each vertex and edges emanating from it are owned by one processor

☐ 1-D partitioning of the adjacency matrix

$$\begin{bmatrix} A_1 \\ A_2 \\ \vdots \\ A_P \end{bmatrix}$$

☐ Edges emanating from vertex v is its edge list = list of vertex indices in row v of adjacency matrix A

1-D Partitioning

- At each level, each processor owns a set F – set of frontier vertices owned by the processor

- Edge lists of vertices in F are merged to form a set of neighboring vertices, N

- Some vertices of N owned by the same processor, while others owned by other processors

- Messages are sent to those processors to add these vertices to their frontier set for the next level

Algorithm 1 Distributed Breadth-First Expansion with 1D Partitioning

1: Initialize $L_{v_s}(v) = \begin{cases} 0, & v = v_s, \text{ where } v_s \text{ is a source} \\ \infty, & \text{otherwise} \end{cases}$

2: **for** $l = 0$ to ∞ **do**
3:     $F \leftarrow \{v \mid L_{v_s}(v) = l\}$, the set of local vertices with level l
4:     **if** $F = \emptyset$ for all processors **then**
5:         Terminate main loop
6:     **end if**
7:     $N \leftarrow \{\text{neighbors of vertices in } F \text{ (not necessarily local)}\}$
8:     **for all** processors q **do**
9:         $N_q \leftarrow \{\text{vertices in } N \text{ owned by processor } q\}$
10:        **Send** N_q to processor q
11:        **Receive** $\bar{N}_q$ from processor q
12:     **end for**
13:     $\bar{N} \leftarrow \bigcup_q \bar{N}_q$ (The $\bar{N}_q$ may overlap)
14:     **for** $v \in \bar{N}$ and $L_{v_s}(v) = \infty$ **do**
15:        $L_{v_s}(v) \leftarrow l + 1$
16:     **end for**
17: **end for**

$L_{vs}(v)$ – level of v, i.e, graph distance from source vs

BFS on GPUs

```
1  bfs_kernel(int curLevel){

2    v = blockIdx.x * blockDim.x + threadIdx.x;
3    if dist[v] == curLevel then
4        forall the n ∈ neighbors(v) do
5            if visited[n] == 0 then
6                dist[n] = dist[v] + 1;
7                visited[n] = 1;
8            end
9        end
10   end
11 }
```

BFS on GPUs

- ☐ One GPU thread for a vertex
- ☐ For each level, a GPU kernel is launched with the number of threads equal to the number of vertices in the graph
- ☐ Only those vertices whose assigned vertices are frontiers will become active
- ☐ Do we need atomics?
- ☐ Severe load imbalance among the treads
- ☐ Scope for improvement