

A Distributed Lattice Boltzmann-based Flow Simulator

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February 22th 2009 / EDi'09

Outline

- 1 Introduction
 - Lattice Boltzmann Flow Simulation
 - Distributed Flow Simulator Challenges
- 2 Dynamic Load Balancing
- 3 Fault-Tolerance
- 4 Concluding Remarks

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Context

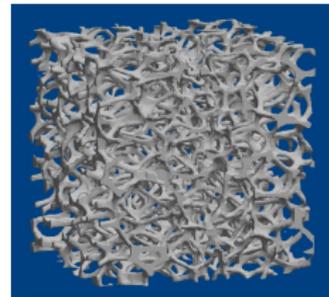
- Flow simulator (LaBoGrid) used by Laboratory of Chemical Engineering of ULg (LGC).
- X-ray tomography gives access to complex geometries (packed beds, metallic foams, ...).
- More accurate description of phenomena occurring in such geometries → Better design, efficiency of related processes.
- Computational Fluid Dynamics (CFD) tools help to reach this goal.

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Why Lattice Boltzmann?

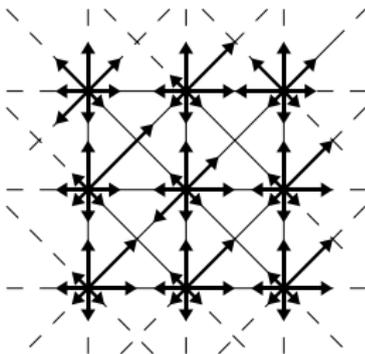
- Tomography produces large matrices of voxels (**solid** in which fluid flows).
- Lattice Boltzmann (LB) methods:
 - can directly use voxel matrices,
 - are easily parallelized,
 - ... but require large computing power and memory.



Solid of metallic foam

Lattice Boltzmann methods

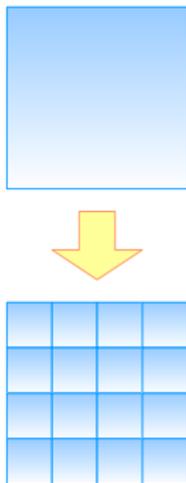
- Fluid = fictive particles moving and colliding on a **lattice** (= regular mesh, space representation).
- State of lattice point = amount of particles moving in fixed directions.
- State at time $t + 1 = f(\text{state at time } t, \text{state of neighbors at time } t)$.



Arrow size = amount of particles moving in a direction.

Distributed Lattice Boltzmann Code I

- Lattice (and solid) is decomposed into **sub-lattices** (sub-solids) of same sizes.



*2D Lattice decomposed
into 16 sub-lattices*

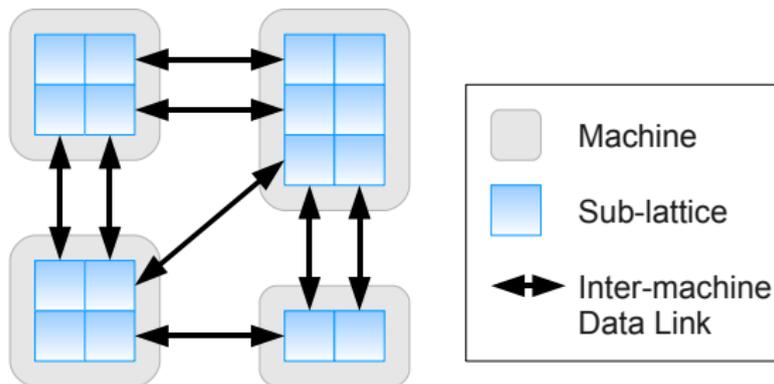
```

i := 0;
do  i < iterations →
    "Send outgoing data";
    "Receive incoming data";
    "Update sub-lattice points";
    i := i + 1
od
  
```

Sub-lattice update Code

Distributed Lattice Boltzmann Code II

- Sub-lattices can be deployed on different machines.
- **Data link** between sub-lattice A and sub-lattice B = outgoing data of A are incoming data of B.



Distributed LB Code Representation

- Distributed LB code can be represented by a graph.
- Node = Sub-lattice (and associated sub-solid).
- Edge = Data link between sub-lattices.
- This graph is called **Model Graph**.

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Environnement : Dynamic Heterogeneous Cluster

- Heterogeneous software environnements
→ Java
- Heterogeneous computing power
→ Load Balancing
- Limited reliability
→ Fault Tolerance
- Available computing power fluctuations (background load)
→ Dynamic Load Balancing (DLB)

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Definitions

A Machine Graph

- Represents a cluster
- Node = Machine
- Edge = Network link
- Nodes are labelled with computing power

Mapping graph $G_1 = (N, E)$ onto graph $G_2 = (M, F)$

- Partition $P(G_1)$ with $|P(G_1)| = |M|$
- $i \in [1..|M|] : P_i(G_1) \rightarrow M_i$

Load Balancing

- Optimization problem:
 - Map Model Graph M onto Machine Graph R
 - Mapping **minimizes execution time**
 - $|P_i(M)| \propto$ power of machine i .
 - **Edge cut** is minimized \rightarrow network data transfers minimized
- Static load balancing: load balancing is done ones for all **before execution**.
- Existing tools: JOSTLE, SCOTCH, METIS, ...

Dynamic Load Balancing Requirements

- Load balancing can occur several times during execution.
- Additional requirements, mapping must be:
 - Calculated fast (\rightarrow parallel/distributed algorithm).
 - Incremental.
 - Calculated in parallel of simulation execution.
- Another additional requirement: **no centralized architecture.**

Existing Methods

- **Iterative** optimization algorithms.
- Optimal workload estimation for each machine (size of Model Graph partition) based on **local information** (workload estimation of subset of available machines).
- After convergence, **migration** of work (Model Graph nodes) from overloaded to underloaded machines.

+

- Easy to implement
- Decentralized

-

- Slow convergence

Proposed Approach

Enhanced Tree Walking Algorithm (Shu'96):

- 1 Machine Graph is organized into a **tree**.
- 2 Bottom-up (leaves \rightarrow root) message propagation to **compute optimal workload**.
- 3 Top-down (root \rightarrow leaves) message propagation to **schedule work migration**.
- 4 Local **partition refinements** (modified KL (Kernighan-Lin'90) refinement with neighboring partitions).

Remarks

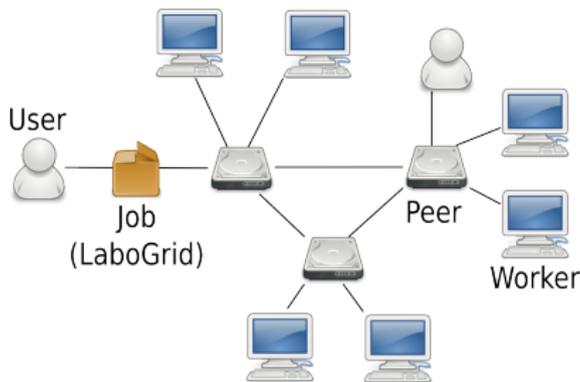
- No iterative optimization $\rightarrow O(10^3)$ less messages, **no slow convergence**.
- Distributed and scalable.
- Incremental.

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LBG-Square

- Joint work with colleague Cyril Briquet (ULg).
- **Combination** of Flow Simulator (LaBoGrid) with a P2P Grid Middleware (LBG).
- LaBoGrid = Job of LBG.



Lightweight Bartering Grid (LBG)

- Peer schedules tasks on its workers (→ peer schedules **local tasks**).
- 1 Task / Worker.
- Tasks can be submitted by peer A to peer B to accelerate job's execution (→ peer B schedules **remote tasks**).
- When peer schedules local tasks, remote tasks' execution can be interrupted (**priority to local task**)
⇒ Tasks preemption

Problem:

- Task interruption (=failure) → flow simulation cannot continue!
- A solution: Checkpoint/restart mechanism.

Checkpoint/restart mechanism

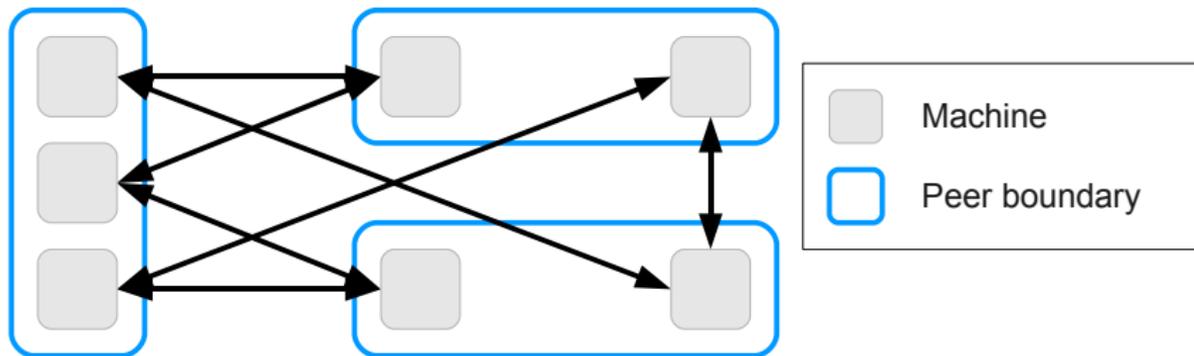
- 1 Each task **saves** its state on a regular time basis (checkpoint).
- 2 In case of **failure**, all tasks **restore** their previous saved state (roll-back).
- 3 The states associated to lost tasks are restored by new or remaining tasks.
- 4 Tasks are **restarted** from restored state.

NB: State of LaBoGrid task = values of associated sub-lattices

Proposed Approach : Distributed Checkpointing

- Task state is replicated on a subset of available machines
→ **Several state replicas per task.**
 - Robust (No single point of failure).
 - Scalable (No central bottleneck).
- State is saved to disk (reduced memory footprint).
- Parameters: number of state replicas and checkpointing frequency.
- Goal : ≥ 1 state copy / task available in case of task(s) interruption(s).
⇒ What state on what machine?

Checkpointing Graph



Properties:

- **Fixed number of outgoing edges** per node (number of state replicas).
- **Load balancing on** the number of **incoming edges** per node.
- Edges should always **cut peer boundaries** (tolerance to preemption).

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Summary and Future Work

- Distributed implementation of a Lattice Boltzmann fluid flow simulator.
- Fast Dynamic Load Balancing.
- Distributed checkpointing.
- What needs to be done?
 - Integrate described dynamic load balancing method (centralized method currently used).
 - Further decentralization.

Thank you for your attention!

Some figures of these slides use pictures from The Gnome Project.

<http://www.montefiore.ulg.ac.be/~dethier/>