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Generated domain-specific sparse data kernels for high-performance Lattice Boltzmann Methods

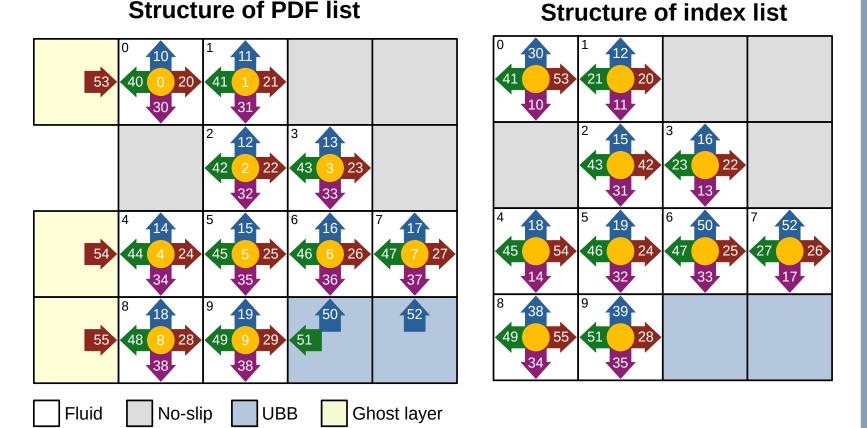
Sparse Lattice Boltzmann Method in waLBerla	Optimization: Communication Hiding and Scalability
barse Data Structure:	Communication Hiding for Sparse Data Structure:



- Indirect addressing: Store only fluid cells of domain
- Save memory and computation time
- No-slip and periodic boundaries handled implicitly
- No extra boundary kernel needed

### **Code Generation:**

- Integrated in the code generation pipeline of *lbmpy/pystencils*
- Generate sparse kernels for CPU and GPU architectures
- Flexible stencils and collision (SRT, TRT, MRT, Cumulants, ...)



PDF list and Index list in memory

30 12 15 16 18 19 50 52 38 39 53 20 42 22 54 24 25 26 55 28

10 11 31 13 14 32 33 17 34 35 41 21 43 23 45 46 47 27 49 51

39 40

Boundaries Ghost layer

. 49 50 51 52 53 54 55

**Pull indices East** 

Pull indices West

Fluid cells

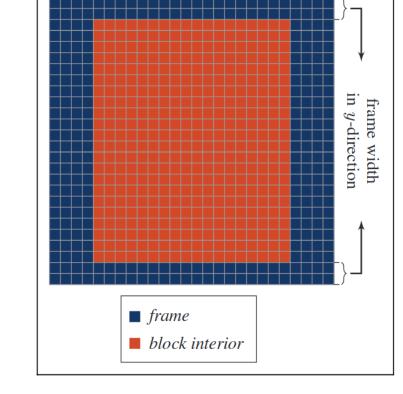
10 ... 19 20 ... 29 30

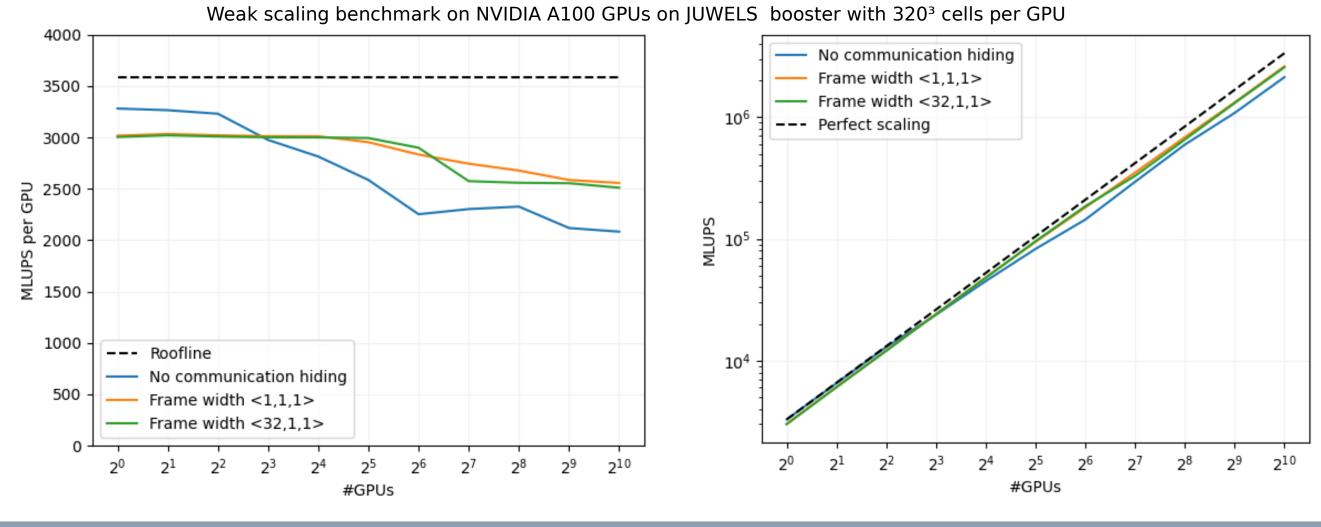
Pull indices North

Pull indices South

### Divide block into interior and frame cells

- Algorithm:
  - 1) Start communication of ghost layers
  - 2) Run kernels on interior cells
  - 3) Wait for communication to finish
  - 4) Run kernels on frame cells
- Hide communication behind inner kernel runs





## Optimization: Hybrid Data Structure

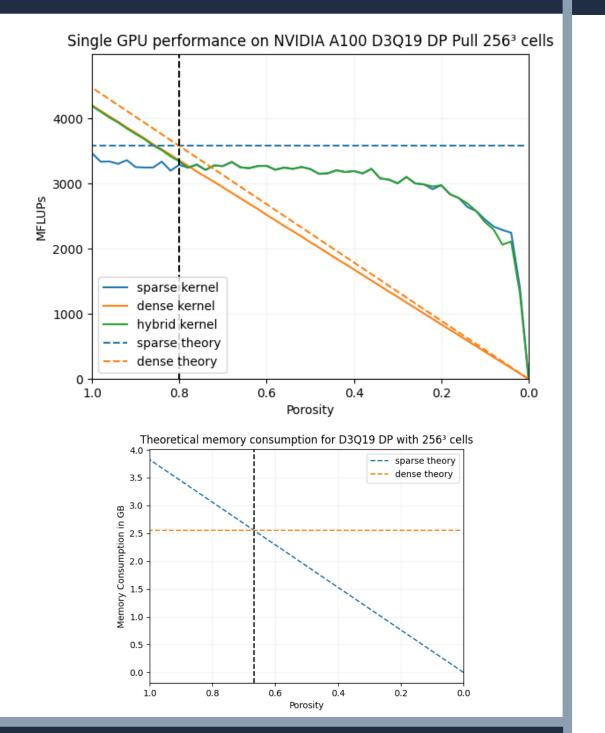
PDF list

Index list

### Sparse LBM Applications

**Domain Decomposition into blocks:** 

- Indirect addressing (sparse) ↔ Direct addressing (dense)
- Decision about data structure per block



#### **LAGOON Test Case:**

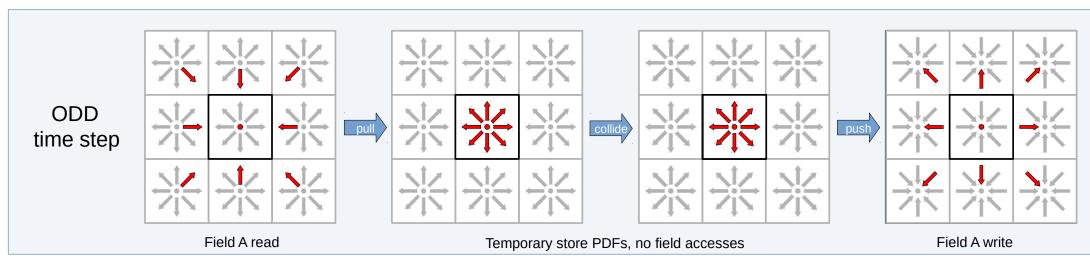
• SCALABLE application: Aircraft landing gear

- Generate sparse and dense kernels based on block porosity
- Porosity = fluid cells / total cells
- Performance superiority of sparse data structure at porosity < 0.8</li>
- Theoretical memory consumption superiority at porosity < 0.66</p>
- Best of both worlds: Hybrid data structure

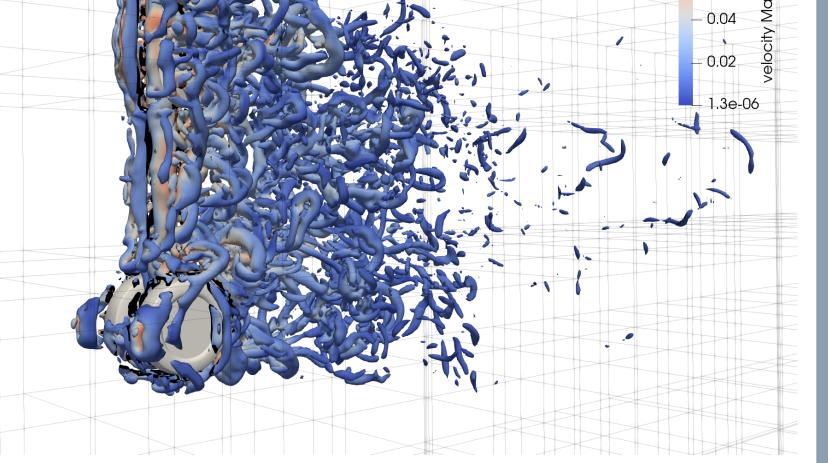
## Optimization: AA Streaming Pattern

### **Theory of in-place Streaming - AA Pattern:**

- Two alternating time steps
- **ODD time step**: PDFs are loaded and stored at the same position
- → For parallel updating no temporary PDF field is needed  $\rightarrow$  ~50% reduced memory consumption
- → Avoid "write allocate" memory access on CPU → Save 33% of memory accesses
- EVEN time step:
- → No streaming, therefore no index list access needed  $\rightarrow$  Save ~10% of memory accesses



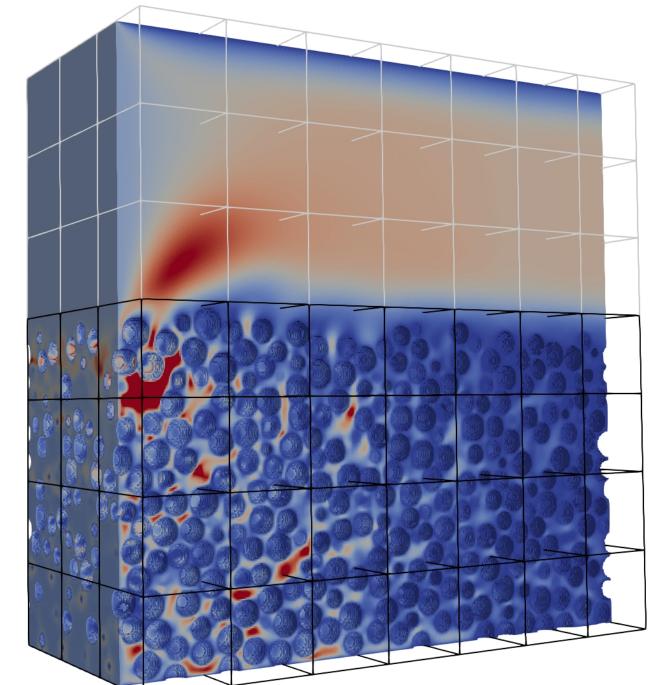
- 720 Block (720 cores on JUWELS Cluser)
- → 188,743,680 cells
- 2 h run time for 3s simulation time
- Reynolds number: 1.59 \* 10<sup>6</sup>
- Sparse data structure only
- But: Low number of boundary cells
- → Test case not suited for sparse LBM



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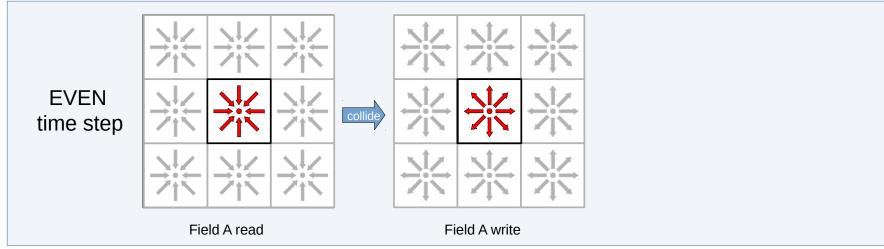
0.06

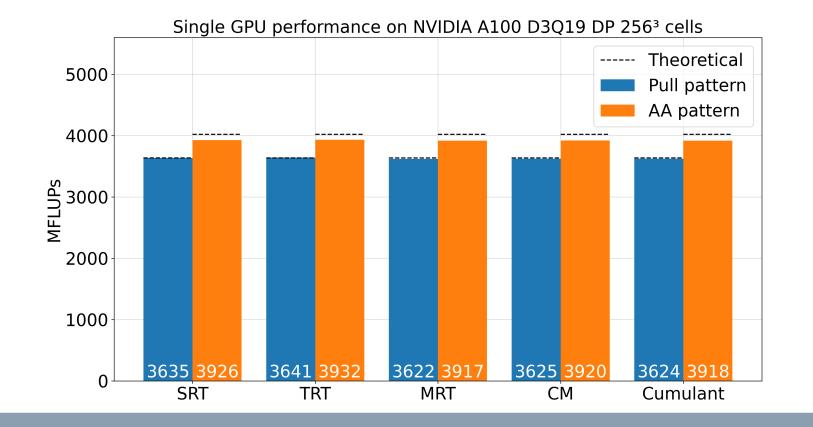
LAGOON test case visualized by Q-criterion. Grey lines are indicating domain decomposition into blocks



### **Artificial Riverbed Example:**

- Porous media and free flow interaction
- Sediment porosity: 0.4 0.6
- Simulation automatically decides per block where to use sparse or dense data structure (porosity threshold < 0.8)</li>
- Well suited case for hybrid data structure
- → Good performance and memory consumption for every part of the domain





#### Riverbed velocity profile. Grey mesh indicates dense blocks, black mesh indicates sparse blocks

#### **Blood Vessels:**

- Complex geometry case with high number of small blocks
- Exclude all blocks without fluid cells from domain
- Remaining blocks have porosity between 0.01 and 1.0 (Ø 0.15)
- Sparse Lattice Boltzmann Method very worth
- Outlook: Balancing workload over processors based on block porosity

