

A block-structured parallel adaptive Lattice-Boltzmann method for rotating geometries

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Outline

Introduction

AMROC software

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Adaptive LBM

- Lattice Boltzmann method
- Structured adaptive mesh refinement
- Verification
- Performance assessment
- Complex geometry consideration

Realistic computations

- Static geometries
- Simulation of wind turbines

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Conclusions

- Things to address

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AMROC

- ▶ Cartesian adaptive fluid solver framework for explicit finite volume methods. Implements for instance Berger-Collela-type AMR.
- ▶ Many shock-capturing methods (MUSCL, (hybrid) WENO, etc.) implemented for complex flux functions.

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- ▶ Used to drive Virtual Test Facility (VTF) FSI software.
- ▶ Targets strongly driven problems (shocks, blast, detonations)
- ▶ Geometry embedding via ghost fluid techniques and level set functions. Distance computation with CPT algorithm [Mauch, 2000].

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- ▶ Geometry embedding via ghost fluid techniques and level set functions. Distance computation with CPT algorithm [Mauch, 2000].
- ▶ ~ 430,000 LOC in C++, C, Fortran-77, Fortran-90.
- ▶ Version V2.0 at <http://www.cacr.caltech.edu/asc>. V1.1 (no complex boundaries) still at <http://amroc.sourceforge.net>.
- ▶ Version used here V3.0 with significantly enhanced parallelization (V2.1 not released).
- ▶ Papers: [Deiterding, 2011, Deiterding and Wood, 2013, Deiterding et al., 2009, Deiterding et al., 2007, Deiterding et al., 2006] and at <http://www.rdeiterding.de>

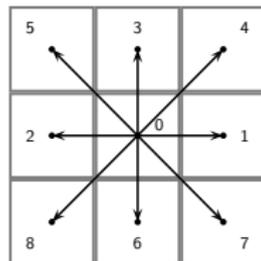
Lattice Boltzmann method

Boltzmann equation: $\partial_t f + \mathbf{u} \cdot \nabla f = \omega(f^{eq} - f)$

Two-dimensional LBM for weakly compressible flows

Formulated on FV grids! (\rightarrow boundary conditions!)

$$\rho(\mathbf{x}, t) = \sum_{\alpha=0}^8 f_{\alpha}(\mathbf{x}, t), \quad \rho(\mathbf{x}, t)u_i(\mathbf{x}, t) = \sum_{\alpha=0}^8 \mathbf{e}_{\alpha i} f_{\alpha}(\mathbf{x}, t)$$



cf. [Hähnel, 2004]

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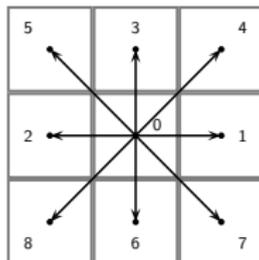
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1.) Transport step \mathcal{T} : $\tilde{f}_{\alpha}(\mathbf{x} + \mathbf{e}_{\alpha}\Delta t, t + \Delta t) = f_{\alpha}(\mathbf{x}, t)$



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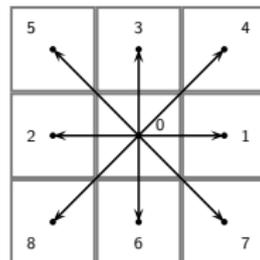
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$$f_{\alpha}(\cdot, t + \Delta t) = \tilde{f}_{\alpha}(\cdot, t + \Delta t) + \omega\Delta t \left(\tilde{f}_{\alpha}^{eq}(\cdot, t + \Delta t) - \tilde{f}_{\alpha}(\cdot, t + \Delta t) \right)$$



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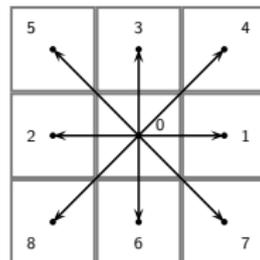
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with equilibrium function

$$f_{\alpha}^{eq}(\rho, \mathbf{u}) = \rho t_{\alpha} \left[1 + \frac{\mathbf{e}_{\alpha} \mathbf{u}}{c_s^2} + \frac{(\mathbf{e}_{\alpha} \mathbf{u})^2}{2c_s^4} - \frac{\mathbf{u}^2}{2c_s^4} \right]$$

mit $t_{\alpha} = \frac{1}{9} \left\{ 4, 1, 1, 1, \frac{1}{4}, \frac{1}{4}, 1, \frac{1}{4}, \frac{1}{4} \right\}$



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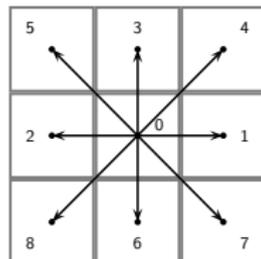
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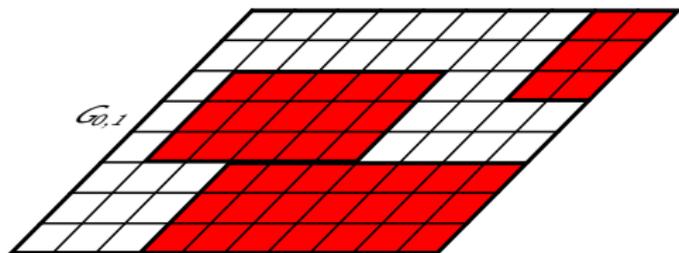
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Lattice speed of sound: $c_s = \frac{1}{\sqrt{3}} \frac{\Delta x}{\Delta t}$, pressure $p = \sum_{\alpha} f_{\alpha}^{eq} c_s^2 = \rho c_s^2 = \rho RT$

Collision frequency vs. kinematic viscosity: $\omega = \frac{c_s^2}{\nu + \Delta t c_s^2 / 2}$ cf. [Hähnel, 2004]

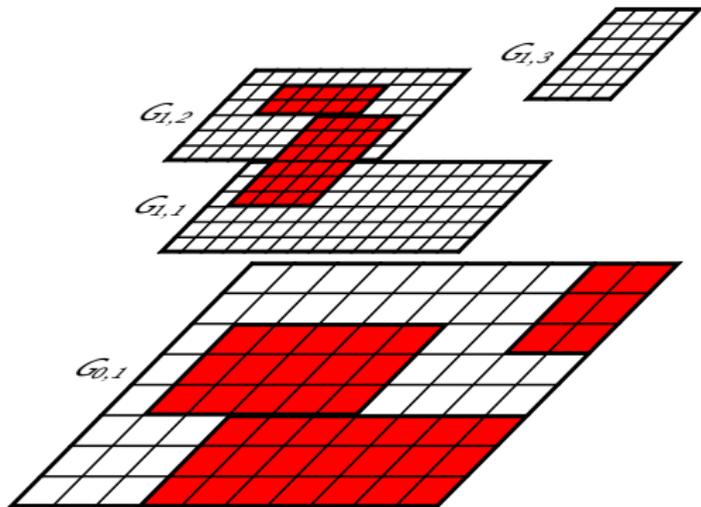
Block-structured adaptive mesh refinement (SAMR)

- ▶ Refined blocks overlay coarser ones



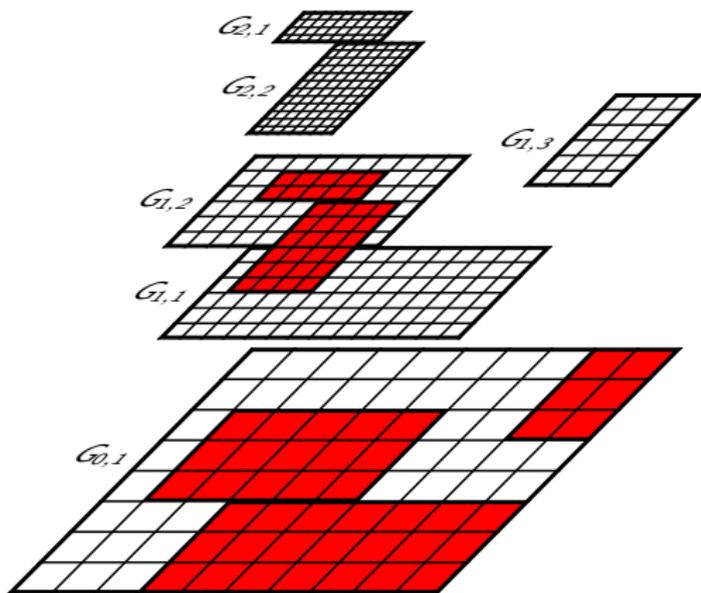
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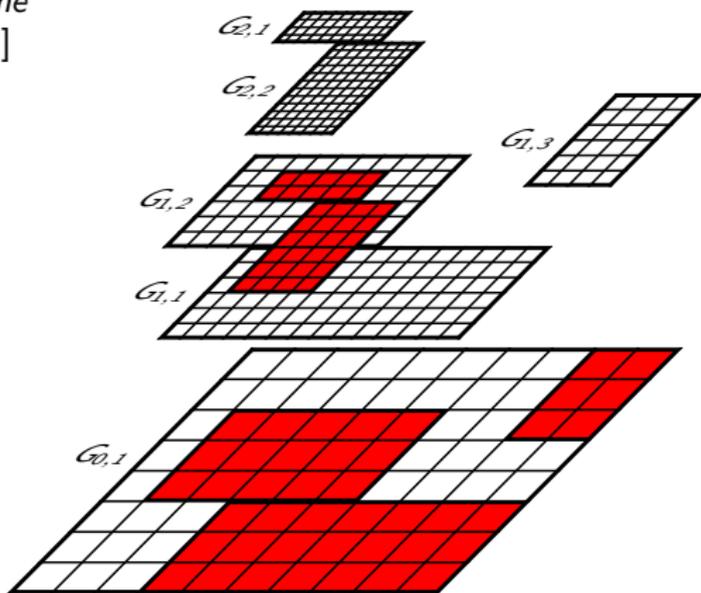
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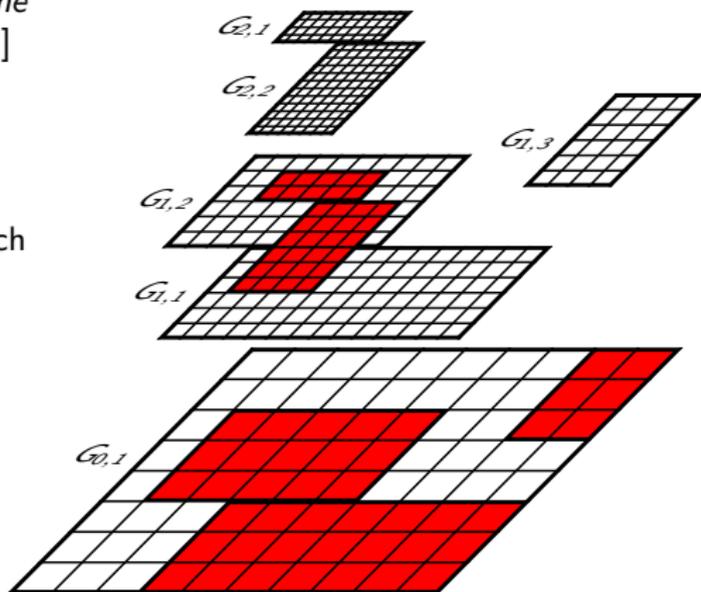
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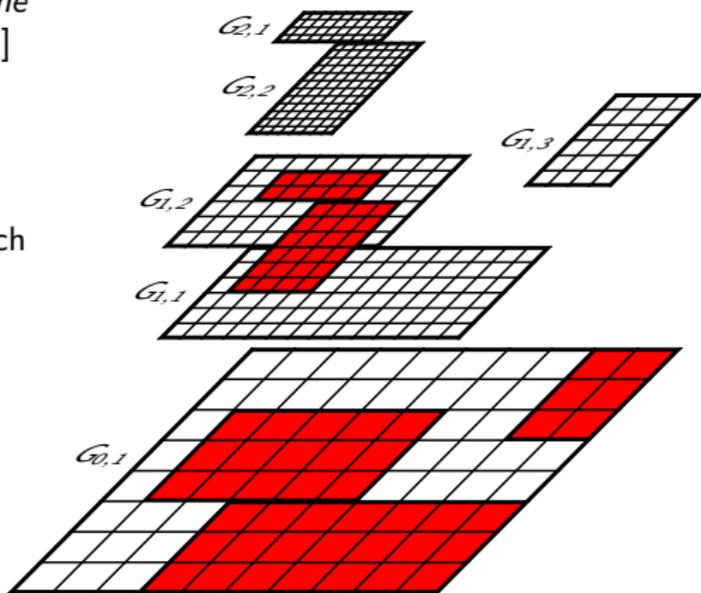
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- + Most efficient LBM implementation with patch-wise for-loops



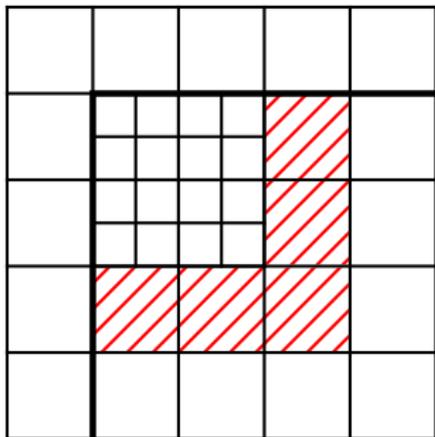
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- + Most efficient LBM implementation with patch-wise for-loops
- + Cache efficient
- ▶ Spatial interpolation and averaging can be used unaltered
- Cluster-algorithm necessary



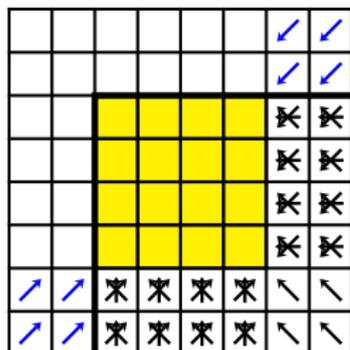
Adaptive LBM

1. Complete update on coarse grid: $f_{\alpha}^{C,n+1} := \mathcal{CT}(f_{\alpha}^{C,n})$



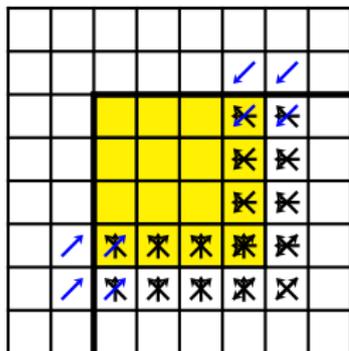
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 $f_{\alpha,in}^{f,n}$

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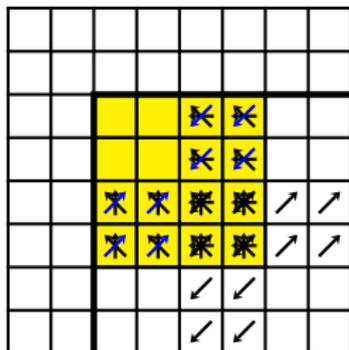
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$$\tilde{f}_{\alpha,in}^{f,n}$$

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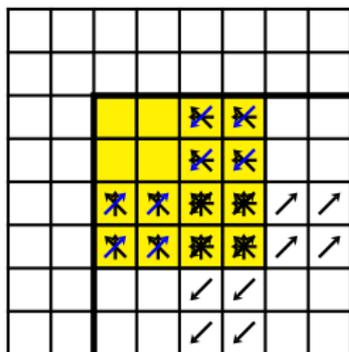
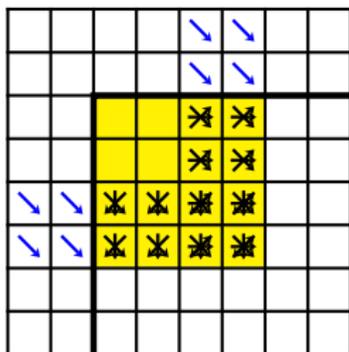
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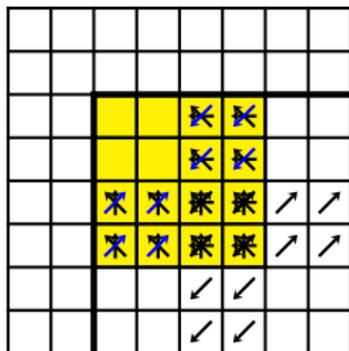
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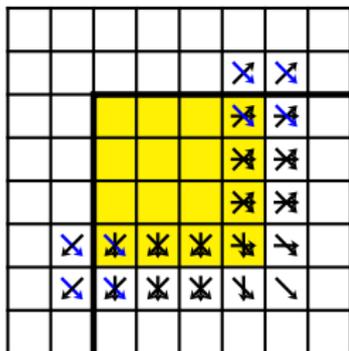
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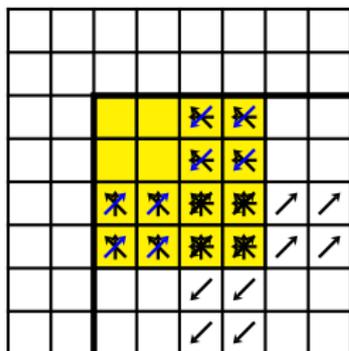


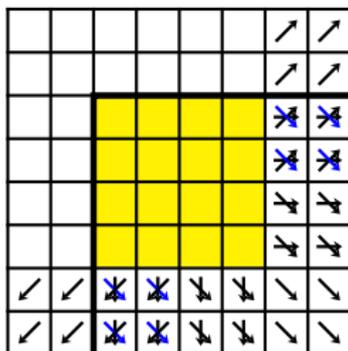
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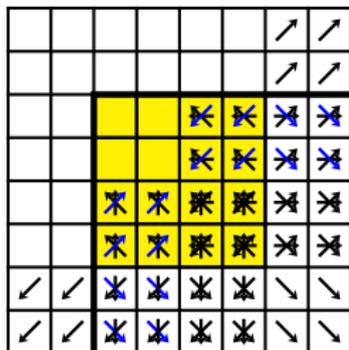


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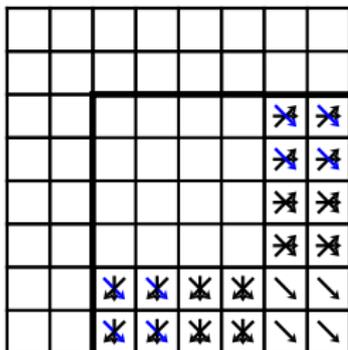
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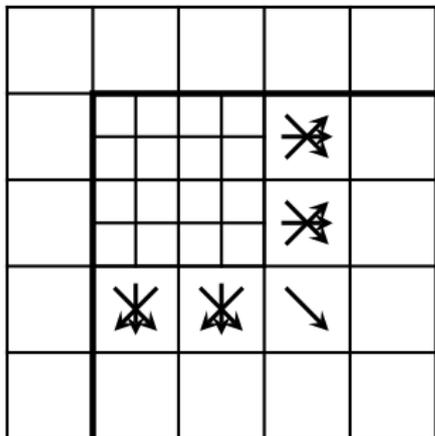


$$\tilde{f}_{\alpha,out}^{f,n+1/2}, \tilde{f}_{\alpha,out}^{f,n}$$

5. Average $\tilde{f}_{\alpha,out}^{f,n+1/2}$ (inner halo layer), $\tilde{f}_{\alpha,out}^{f,n}$ (outer halo layer) to obtain $\tilde{f}_{\alpha,out}^{C,n}$.

Adaptive LBM

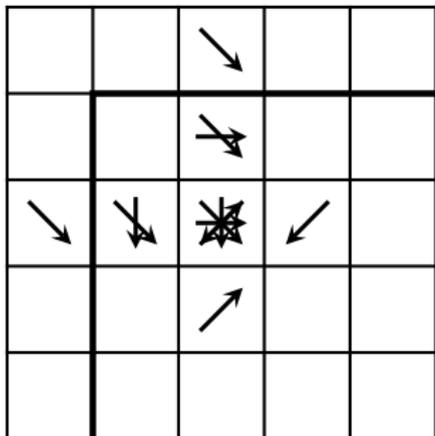
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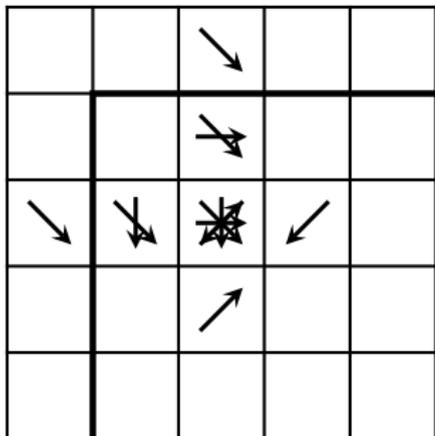
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5. Average $\tilde{f}_{\alpha,out}^{f,n+1/2}$ (inner halo layer), $\tilde{f}_{\alpha,out}^{f,n}$ (outer halo layer) to obtain $\tilde{f}_{\alpha,out}^{C,n}$.
6. Revert transport into halos:
 $\tilde{f}_{\alpha,out}^{C,n} := \mathcal{T}^{-1}(\tilde{f}_{\alpha,out}^{C,n})$

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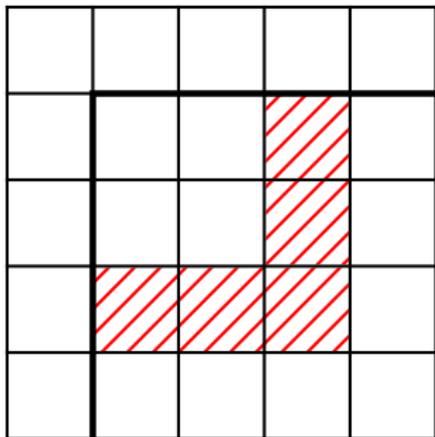
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 $\tilde{f}_{\alpha,out}^{C,n} := \mathcal{T}^{-1}(\tilde{f}_{\alpha,out}^{C,n})$
7. Parallel synchronization of $f_{\alpha}^{C,n}$, $\tilde{f}_{\alpha,out}^{C,n}$

Adaptive LBM

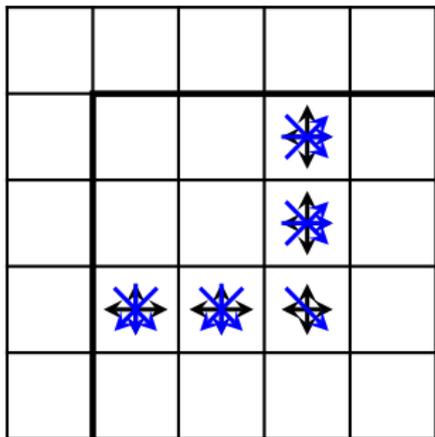
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3. $\tilde{f}_{\alpha}^{f,n} := \mathcal{T}(f_{\alpha}^{f,n})$ on whole fine mesh. $f_{\alpha}^{f,n+1/2} := \mathcal{C}(\tilde{f}_{\alpha}^{f,n})$ in interior.
4. $\tilde{f}_{\alpha}^{f,n+1/2} := \mathcal{T}(f_{\alpha}^{f,n+1/2})$ on whole fine mesh. $f_{\alpha}^{f,n+1} := \mathcal{C}(\tilde{f}_{\alpha}^{f,n+1/2})$ in interior.



5. Average $\tilde{f}_{\alpha,out}^{f,n+1/2}$ (inner halo layer), $\tilde{f}_{\alpha,out}^{f,n}$ (outer halo layer) to obtain $\tilde{f}_{\alpha,out}^{C,n}$.
6. Revert transport into halos:
 $\tilde{f}_{\alpha,out}^{C,n} := \mathcal{T}^{-1}(\tilde{f}_{\alpha,out}^{C,n})$
7. Parallel synchronization of $f_{\alpha}^{C,n}$, $\tilde{f}_{\alpha,out}^{C,n}$
8. Cell-wise update where correction is needed:
 $f_{\alpha}^{C,n+1} := \mathcal{CT}(f_{\alpha}^{C,n}, \tilde{f}_{\alpha,out}^{C,n})$

Adaptive LBM

1. Complete update on coarse grid: $f_{\alpha}^{C,n+1} := \mathcal{CT}(f_{\alpha}^{C,n})$
2. Interpolate $f_{\alpha,in}^{C,n}$ onto $f_{\alpha,in}^{f,n}$ to fill fine halos. Set physical boundary conditions.
3. $\tilde{f}_{\alpha}^{f,n} := \mathcal{T}(f_{\alpha}^{f,n})$ on whole fine mesh. $f_{\alpha}^{f,n+1/2} := \mathcal{C}(\tilde{f}_{\alpha}^{f,n})$ in interior.
4. $\tilde{f}_{\alpha}^{f,n+1/2} := \mathcal{T}(f_{\alpha}^{f,n+1/2})$ on whole fine mesh. $f_{\alpha}^{f,n+1} := \mathcal{C}(\tilde{f}_{\alpha}^{f,n+1/2})$ in interior.

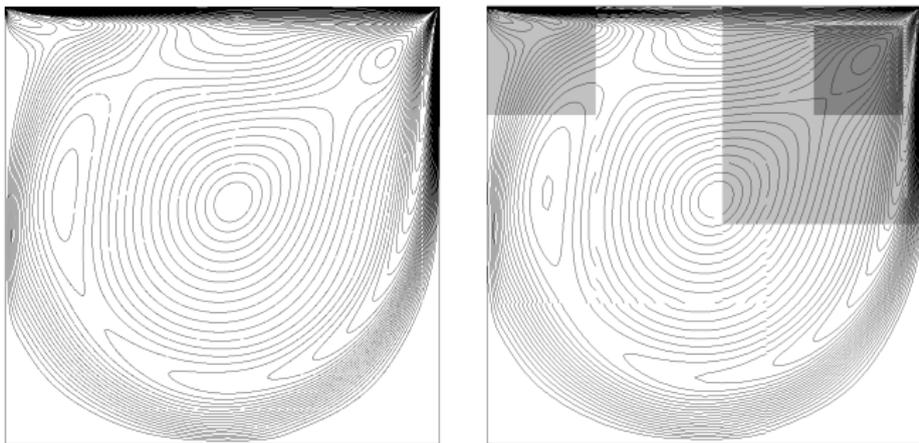


5. Average $\tilde{f}_{\alpha,out}^{f,n+1/2}$ (inner halo layer), $\tilde{f}_{\alpha,out}^{f,n}$ (outer halo layer) to obtain $\tilde{f}_{\alpha,out}^{C,n}$.
6. Revert transport into halos:
 $\bar{f}_{\alpha,out}^{C,n} := \mathcal{T}^{-1}(\tilde{f}_{\alpha,out}^{C,n})$
7. Parallel synchronization of $f_{\alpha}^{C,n}$, $\bar{f}_{\alpha,out}^{C,n}$
8. Cell-wise update where correction is needed:
 $f_{\alpha}^{C,n+1} := \mathcal{CT}(f_{\alpha}^{C,n}, \bar{f}_{\alpha,out}^{C,n})$

Algorithm equivalent to [Chen et al., 2006].

Verification - driven cavity

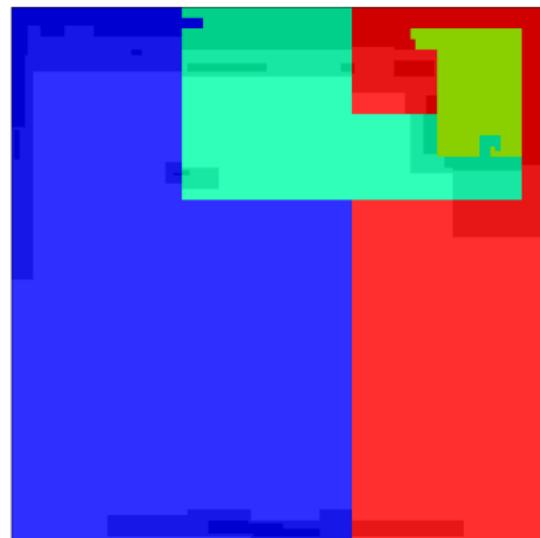
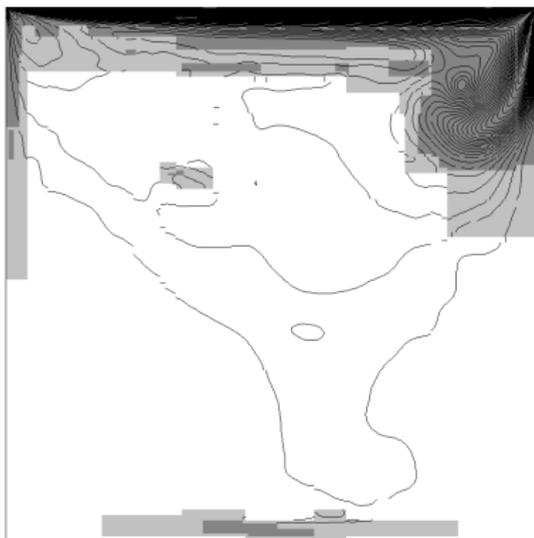
- ▶ $Re = 1500$ in air, $\nu = 1.5 \cdot 10^{-5} \text{ m}^2/\text{s}$, $u = 22.5 \text{ m/s}$.
- ▶ Domain size $1 \text{ mm} \times 1 \text{ mm}$.
- ▶ Reference computation uses 800×800 lattice.
- ▶ 588,898 time steps to $t_e = 5 \cdot 10^{-3} \text{ s}$, $\sim 35 \text{ h CPU}$.
- ▶ Statically adaptive computation uses 100×100 lattice with $r_{1,2} = 2$.
- ▶ 294,452 time steps to $t_e = 5 \cdot 10^{-3} \text{ s}$ on finest level.



Isolines of density. Left: reference, right on refinement at t_e .

Driven cavity - dynamic refinement

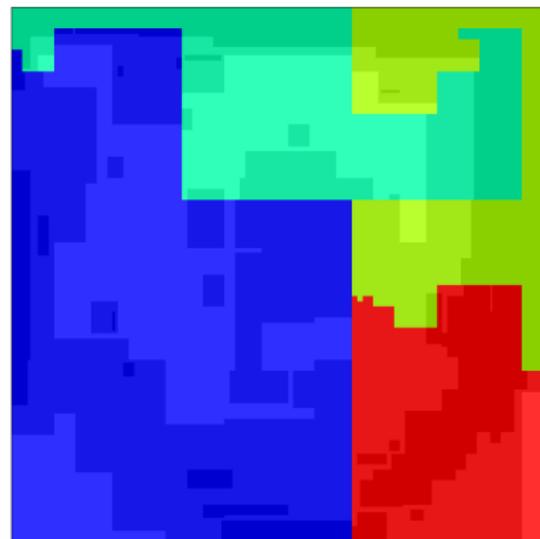
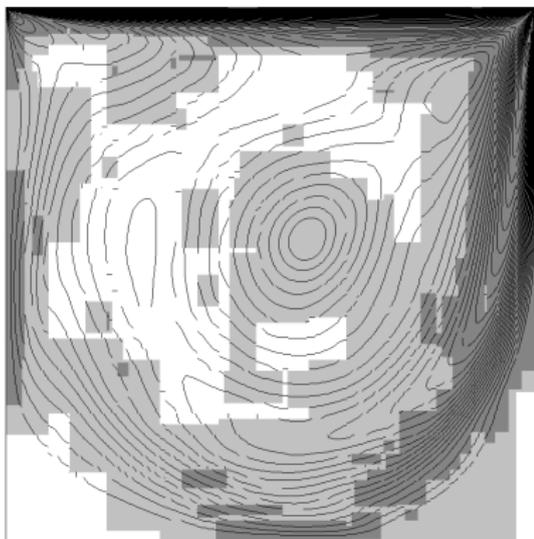
- ▶ Dynamic refinement based on heuristic error estimation of $|\mathbf{u}|$
- ▶ Threshold intentionally chosen to show refinement evolution



Isolines of density on refinement (left), distribution to 4 processors (right).

Driven cavity - dynamic refinement

- ▶ Dynamic refinement based on heuristic error estimation of $|\mathbf{u}|$
- ▶ Threshold intentionally chosen to show refinement evolution



Isolines of density on refinement (left), distribution to 4 processors (right).

Driven cavity - 3d cavity

- ▶ Similar setup as in 2d. No-slip wall everywhere except at lid. $Re = 1000$ in air, $u = 15$ m/s.
- ▶ AMR 64^3 base mesh with $r_{1,2} = 2$. Regridding and repartition only at every 2nd base level step.
- ▶ 95 time steps on coarsest level benchmarked.
- ▶ Uniform grid $256^3 = 16.8 \cdot 10^6$ cells.

Level	Grids	Cells
0	178	262,144
1	668	1,538,912
2	2761	7,842,872

Grid and cells used on 24 cores

Cores	6	12	24	48	96
Time per step	1.82s	0.94s	0.50s	0.28s	0.16s
LBM Update	44.97%	42.83%	39.64%	35.37%	31.10%
Error Estimation	1.37%	1.30%	1.20%	1.07%	0.94%
Regridding	14.59%	14.79%	15.60%	16.75%	19.14%
Fixup	4.18%	3.96%	3.74%	3.42%	3.07%
Interp. Boundaries	9.34%	9.15%	8.30%	7.17%	6.13%
Interp. Regridding	3.53%	3.23%	3.02%	2.73%	2.44%
Sync Boundaries	8.69%	11.20%	14.28%	18.26%	21.07%
Sync Fixup	2.41%	3.41%	4.70%	6.50%	7.99%
Sync Regridding	0.77%	0.72%	0.74%	0.83%	0.99%
Phys. Boundaries	0.69%	0.68%	0.63%	0.56%	0.49%
Clustering	0.55%	0.48%	0.44%	0.40%	0.36%
Misc	8.90%	8.25%	7.72%	6.95%	6.26%

Driven cavity - 3d cavity

- ▶ Intel Xeon-2.67 GHz 6-core (Westmere) dual-processor nodes with Qlogics interconnect
- ▶ Unigrid with 2 ghost cells

Cores	6	12	24	48	96
Time per step	3.44s	1.81s	0.92s	0.47s	0.24s
Par. Efficiency	100.00%	95.04%	93.34%	92.40%	91.38%
LBM Update	74.86%	74.60%	72.48%	70.78%	67.58%
Synchronization	12.42%	13.32%	15.48%	17.35%	20.52%
Phys. Boundary	0.78%	0.74%	0.69%	0.69%	0.65%
Misc	11.94%	11.34%	11.34%	11.18%	11.25%

- ▶ AMR with 2 ghost cells

Cores	6	12	24	48	96
Time per step	1.82s	0.94s	0.50s	0.28s	0.16s
Par. Efficiency	100.00%	96.47%	90.00%	81.68%	73.04%
LBM Update	46.34%	44.13%	40.84%	36.44%	32.04%
Synchronization	11.87%	15.33%	19.72%	25.58%	30.06%
Phys. Boundary	0.69%	0.68%	0.63%	0.56%	0.49%
Regridding	14.59%	14.79%	15.60%	16.75%	19.14%
Interpolation	12.88%	12.38%	11.31%	9.90%	8.57%
Fixup	4.18%	3.96%	3.74%	3.42%	3.07%
Misc	9.45%	8.73%	8.16%	7.36%	6.62%

Driven cavity - 3d cavity

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Synchronization	11.87%	15.33%	19.72%	25.58%	30.06%
Phys. Boundary	0.69%	0.68%	0.63%	0.56%	0.49%
Regridding	14.59%	14.79%	15.60%	16.75%	19.14%
Interpolation	12.88%	12.38%	11.31%	9.90%	8.57%
Fixup	4.18%	3.96%	3.74%	3.42%	3.07%
Misc	9.45%	8.73%	8.16%	7.36%	6.62%

- ▶ Expense for boundary is increased compared to FV methods because the algorithm uses few floating point operations but a large state vector!

Driven cavity - 3d cavity

► Unigrid with 1 ghost cell

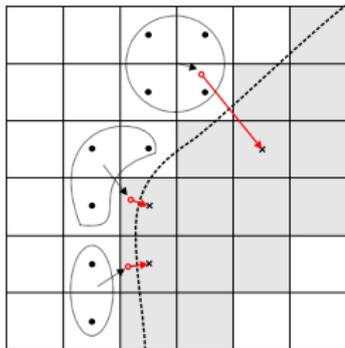
Cores	6	12	24	48	96
Time per step	2.80s	1.46s	0.73s	0.37s	0.18s
Par. Efficiency	100.00%	96.09%	95.33%	95.21%	94.82%
LBM Update	78.05%	77.08%	75.85%	74.50%	71.38%
Synchronization	7.25%	8.67%	10.00%	11.32%	14.35%
Phys. Boundary	0.51%	0.46%	0.45%	0.44%	0.44%
Misc	14.19%	13.79%	13.70%	13.73%	13.83%

► AMR with 4 ghost cells

Cores	6	12	24	48	96
Time per step	3.32s	1.90s	1.21s	0.54s	0.30s
Par. Efficiency	100.00%	87.42%	68.76%	77.02%	68.19%
LBM Update	43.44%	40.93%	31.33%	34.64%	30.11%
Synchronization	14.13%	18.26%	34.73%	25.76%	30.69%
Phys. Boundary	1.03%	0.98%	0.77%	0.86%	0.77%
Regridding	15.53%	16.02%	13.87%	18.72%	20.82%
Interpolation	16.74%	15.71%	11.95%	13.15%	11.51%
Fixup	2.89%	2.60%	2.02%	2.28%	2.03%
Misc	6.22%	5.50%	5.33%	4.59%	4.08%

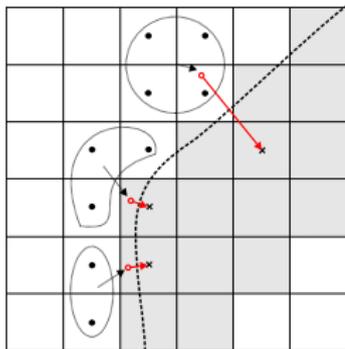
► Basically linear dependency on number of ghost cells used

Level-set method for boundary embedding



- ▶ Implicit boundary representation via distance function φ , normal $\mathbf{n} = \nabla\varphi/|\nabla\varphi|$.
- ▶ Complex boundary moving with local velocity \mathbf{w} , treat interface as moving rigid wall.
- ▶ Construction of macro-values in embedded boundary cells by interpolation / extrapolation.

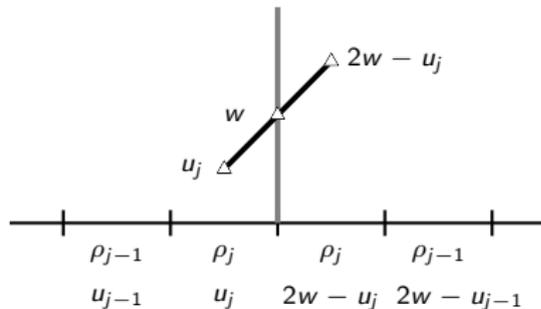
Level-set method for boundary embedding



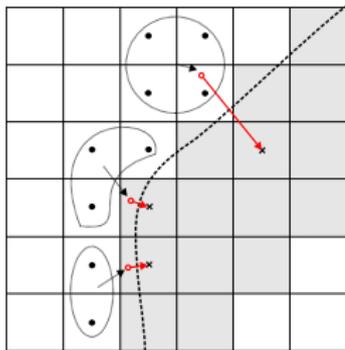
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Interpolate / constant value extrapolate values at

$$\tilde{\mathbf{x}} = \mathbf{x} + 2\varphi\mathbf{n}$$



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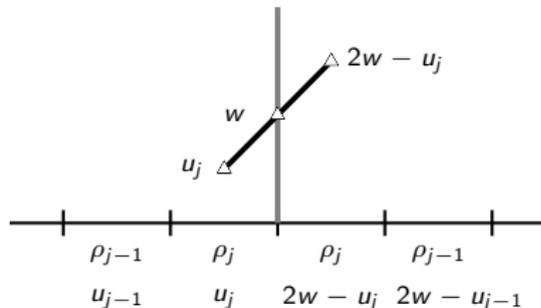
$$\tilde{\mathbf{x}} = \mathbf{x} + 2\varphi\mathbf{n}$$

Macro-velocity in ghost cells:

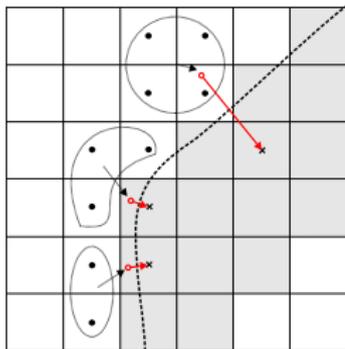
No-slip: $\mathbf{u}' = 2\mathbf{w} - \mathbf{u}$

Slip:

$$\begin{aligned}\mathbf{u}' &= (2\mathbf{w} \cdot \mathbf{n} - \mathbf{u} \cdot \mathbf{n})\mathbf{n} + (\mathbf{u} \cdot \mathbf{t})\mathbf{t} \\ &= 2((\mathbf{w} - \mathbf{u}) \cdot \mathbf{n})\mathbf{n} + \mathbf{u}\end{aligned}$$



Level-set method for boundary embedding



- ▶ Implicit boundary representation via distance function φ , normal $\mathbf{n} = \nabla\varphi/|\nabla\varphi|$.
- ▶ Complex boundary moving with local velocity \mathbf{w} , treat interface as moving rigid wall.
- ▶ Construction of macro-values in embedded boundary cells by interpolation / extrapolation.
- ▶ Then use $f_{\alpha}^{eq}(\rho', \mathbf{u}')$ to construct distributions in embedded ghost cells.
- ▶ 2nd order improvements possible, cf. [Peng and Luo, 2008].

Interpolate / constant value extrapolate values at

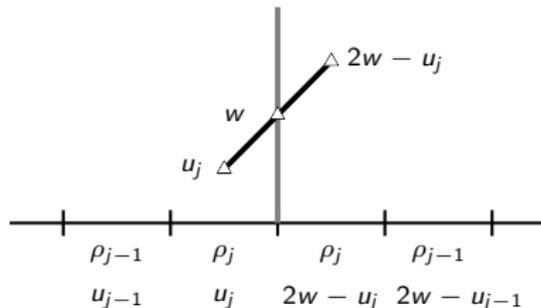
$$\tilde{\mathbf{x}} = \mathbf{x} + 2\varphi\mathbf{n}$$

Macro-velocity in ghost cells:

No-slip: $\mathbf{u}' = 2\mathbf{w} - \mathbf{u}$

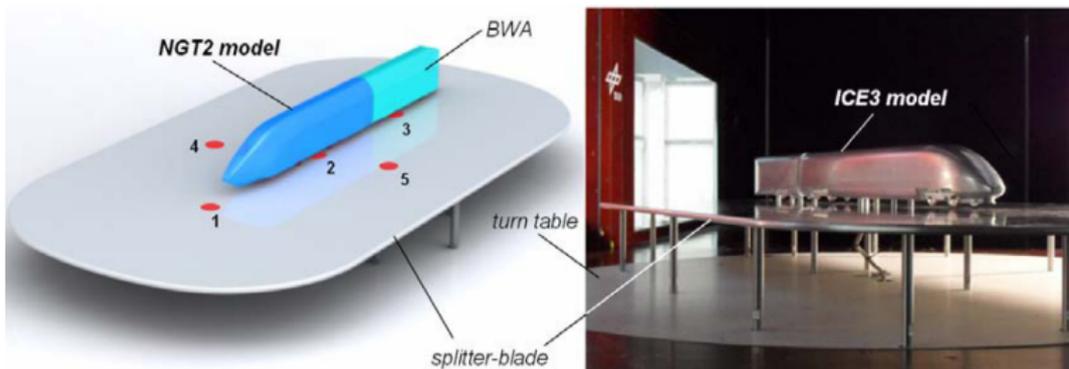
Slip:

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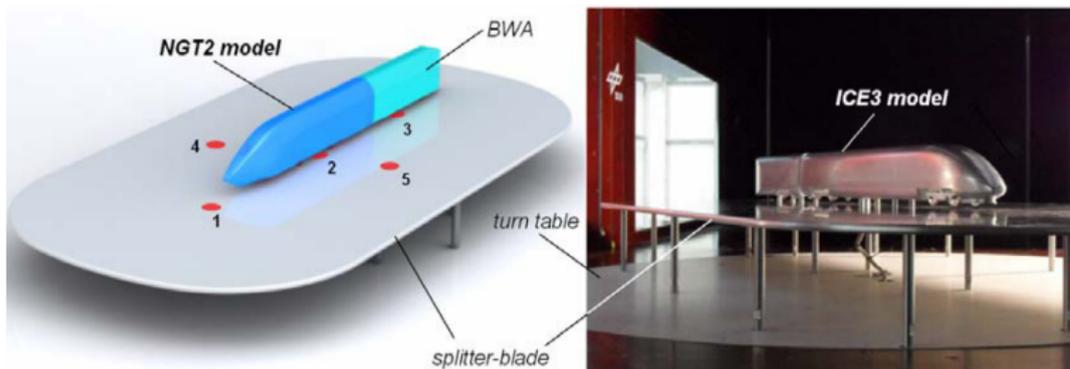
Side-wind investigation for a train model

- ▶ 1:25 train model represented with 74,670 triangles (41,226 front body, 12,398 back body, 21,006 blade)



Side-wind investigation for a train model

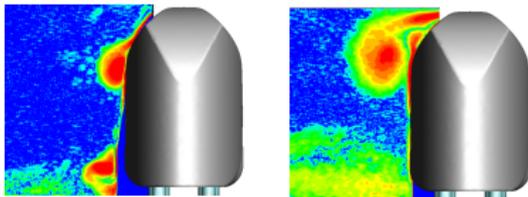
- ▶ 1:25 train model represented with 74,670 triangles (41,226 front body, 12,398 back body, 21,006 blade)
- ▶ Wind tunnel conditions: air at room temperature with 60.25 m/s ($M = 0.18$), $Re = 450,000$
- ▶ Systematic side wind investigation with $0 \geq \beta \geq 30^\circ$ to obtain lift, drag and roll moment coefficients
- ▶ Instationary, turbulent flow conditions make replacing/supplementing experiments with simulations very challenging. Typical DLR problem and good real-world CFD benchmark.



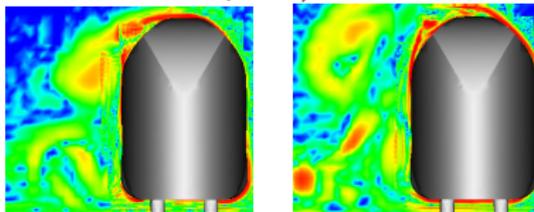
Flow prediction, $Re = 450,000$, $\beta = 30^\circ$

- ▶ Domain $10\text{ m} \times 2.4\text{ m} \times 1.6\text{ m}$
- ▶ Computation started in 3 steps. Full resolution after 5889 coarsest level steps or $t \geq 0.4\text{ s}$
- ▶ ~ 1140 coarsest level steps in 24 h on 96 cores shown above. Overall cost $\sim 4600\text{ h CPU}$.

Experiment (time-averaged)



AMROC-LBM Simulation (instantaneous snapshots)

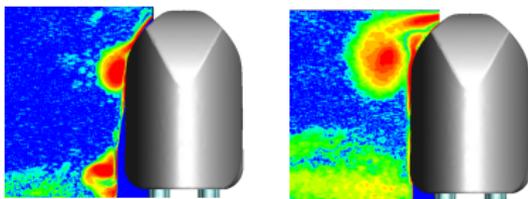


Vorticity component (seen from behind) in axial direction 80 mm and 290 mm away from model tip.

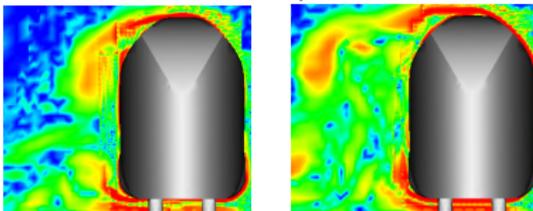
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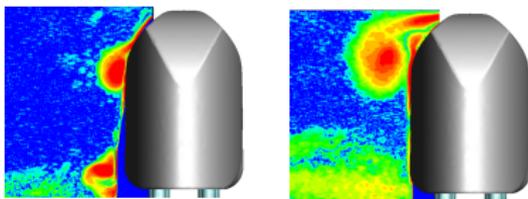


Vorticity component (seen from behind) in axial direction 80 mm and 290 mm away from model tip.

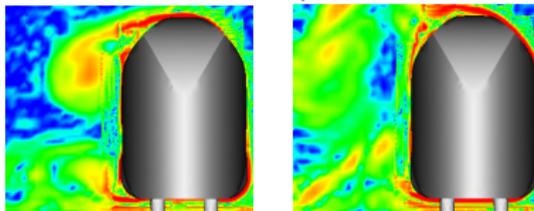
Flow prediction, $Re = 450,000$, $\beta = 30^\circ$

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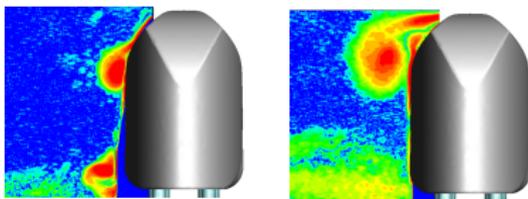


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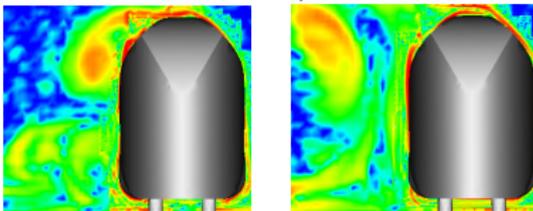
Flow prediction, $Re = 450,000$, $\beta = 30^\circ$

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Experiment (time-averaged)



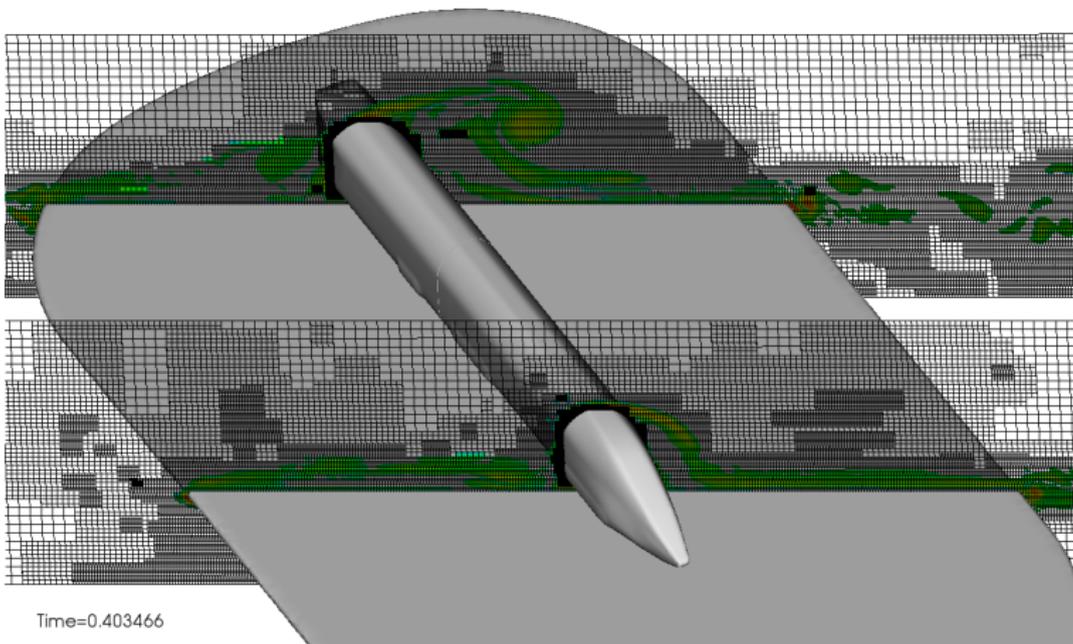
AMROC-LBM Simulation (instantaneous snapshots)



Vorticity component (seen from behind) in axial direction 80 mm and 290 mm away from model tip.

Dynamic mesh adaptation

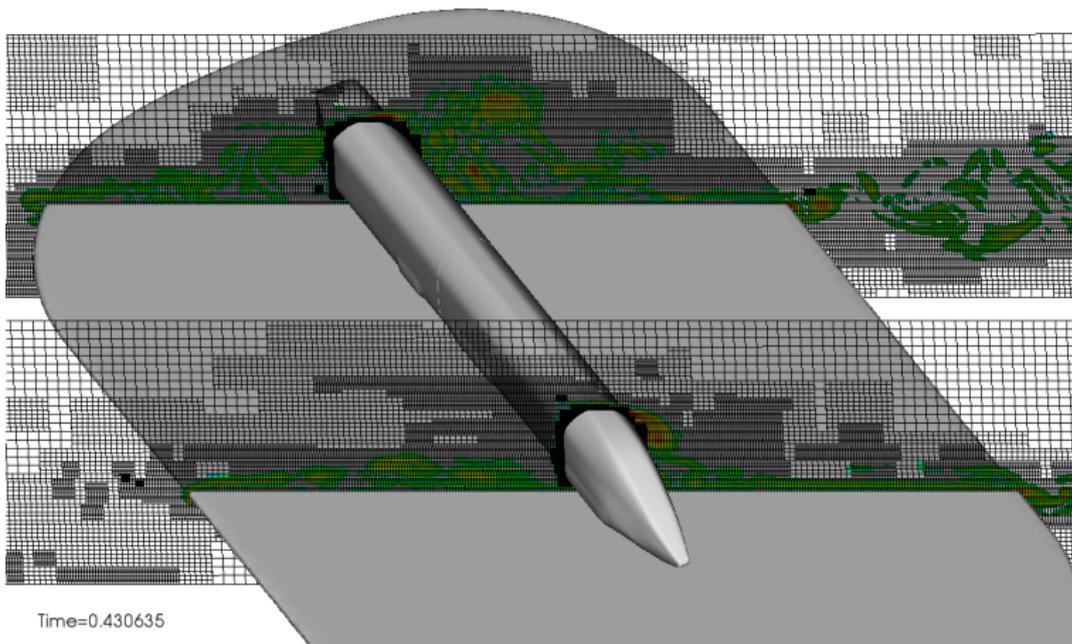
- ▶ Base mesh $500 \times 120 \times 80$ cells, refinement factors 2,2,4.
- ▶ Refinement based on error estimation of $|\mathbf{u}|$ up to second highest level.
- ▶ Highest level reserved to geometry refinement with $\Delta x = 1.25$ mm.



Dynamically adapting mesh. View in wind direction.

Dynamic mesh adaptation

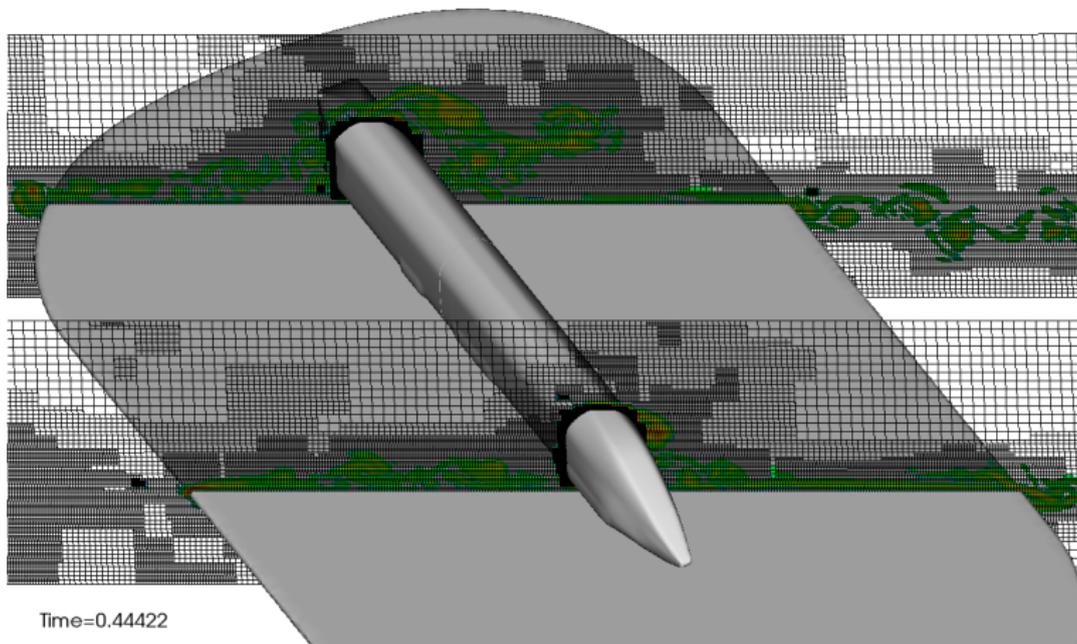
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Dynamic mesh adaptation

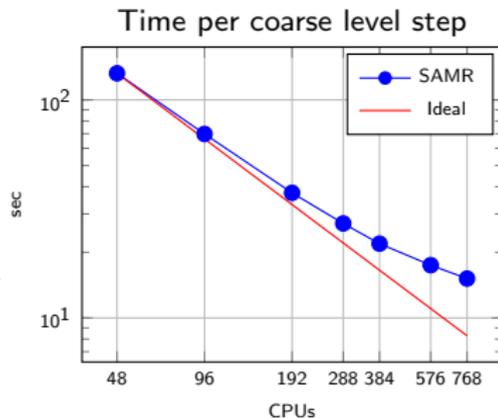
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Dynamically adapting mesh. View in wind direction.

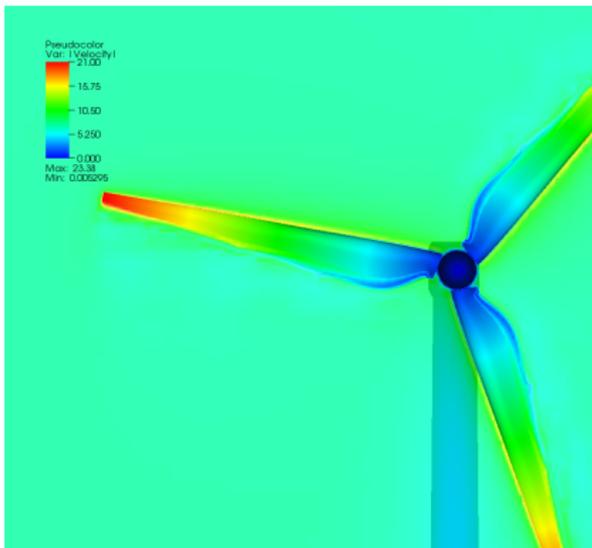
Strong scalability test

- ▶ Computation is restarted from disk checkpoint at $t = 0.526408$ s.
- ▶ Time for initial re-partitioning removed from benchmark.
- ▶ 200 coarse level time steps computed.
- ▶ Regridding and re-partitioning every 2nd level-0 step.
- ▶ Computation starts with 51.8M cells (l3: 10.2M, l2: 15.3M, l1: 21.5M, l0= 4.8M) vs. 19.66 billion (uniform).



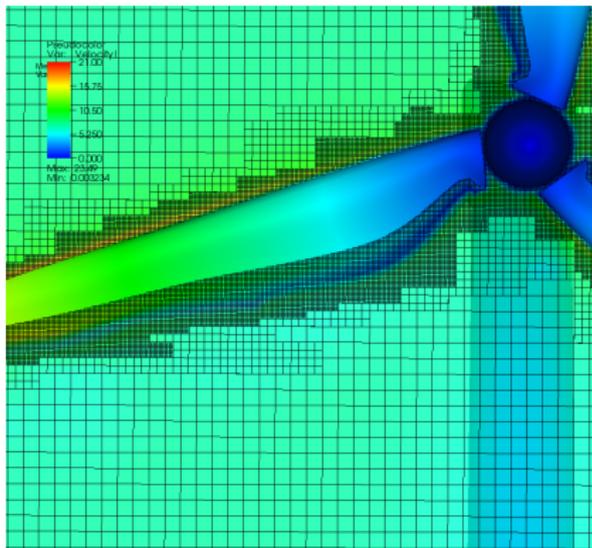
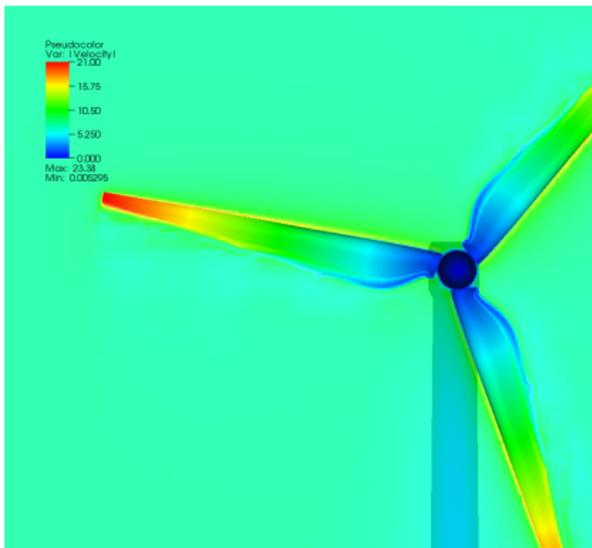
Simulation of a single turbine

- ▶ Geometry from realistic Vestas V27 turbine. Rotor diameter 27 m, tower height ~ 35 m. Ground considered.
- ▶ Prescribed motion of rotor with 15 rpm. Inflow velocity 7 m/s.
- ▶ Simulation domain $200 \text{ m} \times 100 \text{ m} \times 100 \text{ m}$.
- ▶ Base mesh $400 \times 200 \times 200$ cells with refinement factors 2,2,4. Resolution of rotor and tower $\Delta x = 3.125 \text{ cm}$.
- ▶ 141,344 highest level iterations to $t_e = 30 \text{ s}$ computed.



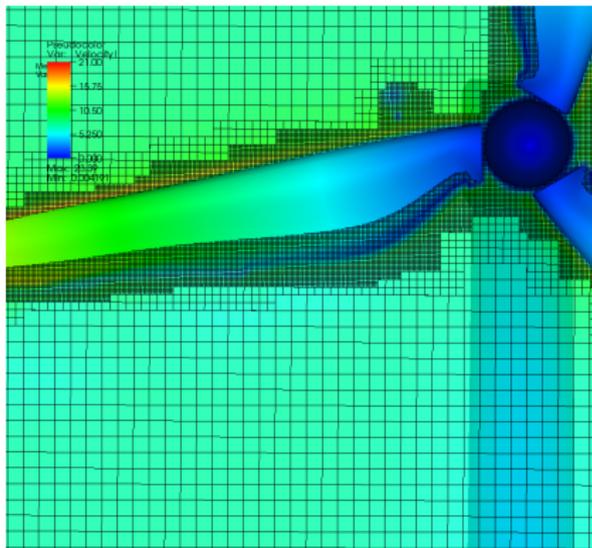
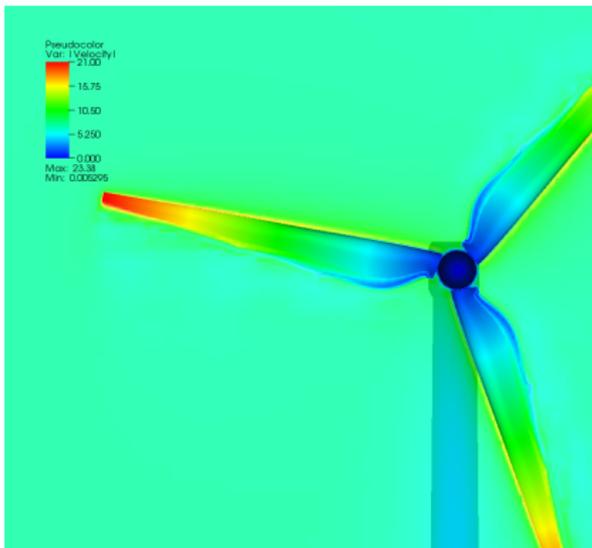
Simulation of a single turbine

- ▶ Geometry from realistic Vestas V27 turbine. Rotor diameter 27 m, tower height ~ 35 m. Ground considered.
- ▶ Prescribed motion of rotor with 15 rpm. Inflow velocity 7 m/s.
- ▶ Simulation domain $200 \text{ m} \times 100 \text{ m} \times 100 \text{ m}$.
- ▶ Base mesh $400 \times 200 \times 200$ cells with refinement factors 2,2,4. Resolution of rotor and tower $\Delta x = 3.125 \text{ cm}$.
- ▶ 141,344 highest level iterations to $t_e = 30 \text{ s}$ computed.

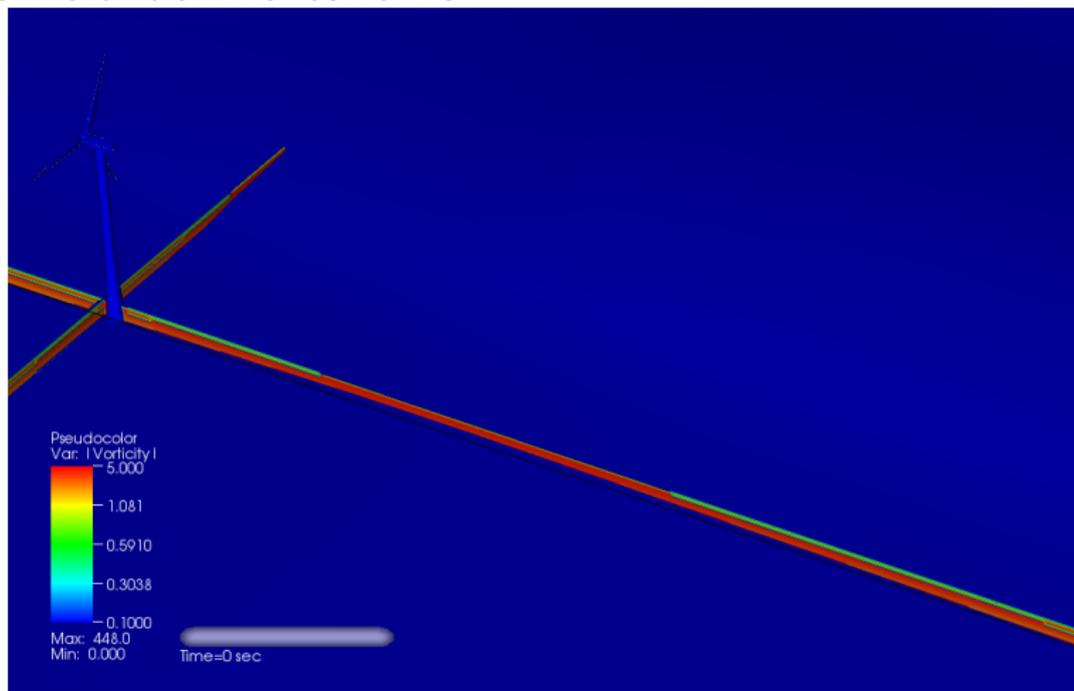


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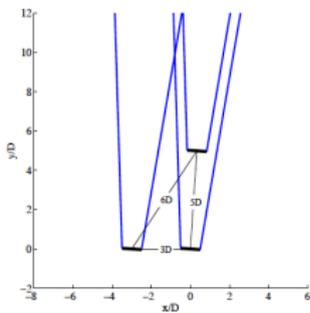
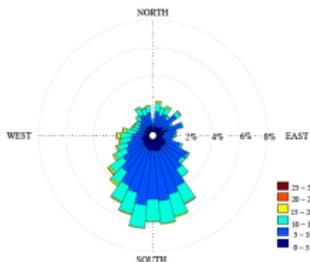
Wake field behind turbine



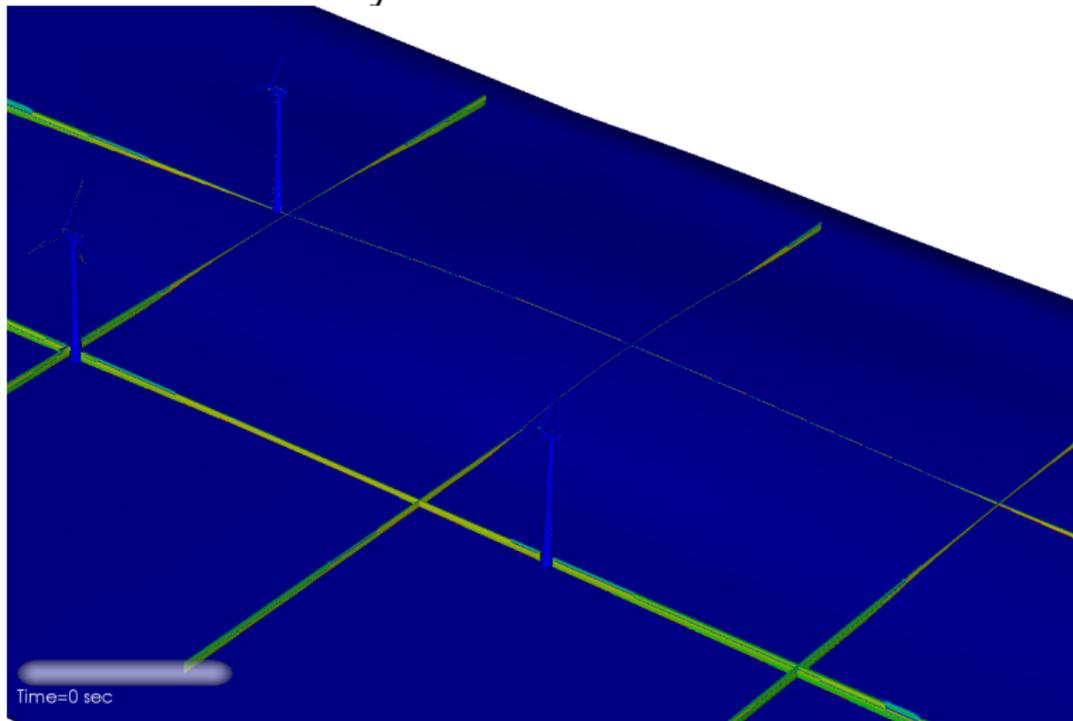
- ▶ Simulation on 96 cores Intel Xeon-Westmere. $\sim 10,400$ h CPU.
- ▶ Error estimation in $|\mathbf{u}|$ refines wake up to level 1 ($\Delta x = 25$ cm).
- ▶ Rotation starts at $t = 4$ s.

Simulation of the SWIFT array

- ▶ Three Vestas V27 turbines. 225 kW power generation at wind speeds 14 to 25 m/s (then cut-off).
- ▶ Prescribed motion of rotor with 15 rpm. Inflow velocity 7 m/s (power generation 52.5 kW).
- ▶ Simulation domain $488 \text{ m} \times 240 \text{ m} \times 100 \text{ m}$.
- ▶ Base mesh $448 \times 240 \times 100$ cells with refinement factors 2,2,2. Resolution of rotor and tower $\Delta x = 12.5 \text{ cm}$.
- ▶ 47,120 highest level iterations to $t_e = 40 \text{ s}$ computed.



Wakes in SWIFT array



- ▶ Simulation on 288 cores Intel Xeon-Westmere. $\sim 140,000$ h CPU.
- ▶ Refinement of wake up to level 2 ($\Delta x = 25$ cm).
- ▶ Rotation starts at $t = 4$ s, full refinement at $t = 8$ s to avoid refining initial acoustic waves.

Conclusions

- ▶ Developed and demonstrated a first version of a block-based dynamically adaptive LBM for real-world CFD with moving boundaries.
- ▶ Reuse of templated AMROC classes from previous finite volume methods already provides robust real-world capabilities.

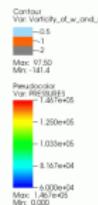
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Conclusions

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- ▶ Reuse of templated AMROC classes from previous finite volume methods already provides robust real-world capabilities.
- ▶ Improve refinement criteria (e.g., vorticity-based) to capture wake fields reliably.
- ▶ Performance for moderate core count is reasonable, some improvements for larger core count still desirable.
 - ▶ Reduce communication width to a single halo layer.
 - ▶ Consider workload due to embedded boundary method in partitioning algorithm.
 - ▶ Allow other than rigorous domain decomposition.
- ▶ Use system for understanding turbine-turbine interactions.
- ▶ Realistic turbine model with dynamic pitch angle, nacelle rotation, etc. under development.

NREL 5 MW turbine



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