# Comparison of Adaptive and Uniform 2D Galerkin Simulations

Andreas Müller, Michal Kopera, Simone Marras, Francis X. Giraldo

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- Motivation
- Results: what do we gain by using AMR?
- Next steps



- **dynamical core** inside the Navy's next generation weather prediction system NEPTUNE
- unified across numerics (contains Continuous and Discontinuous Galerkin methods)
- **unified across applications** (regional and global modeling)
- 3D, DG, MPI: strong scaling for explicit time integration (tested up to 32000 CPUs)
- 2D, serial: allows dynamic AMR

motivation

results





### **NUMA** Collaborators

### **Numerical Methods and Moisture**

Francis X. Giraldo, Applied Math, NPS Michal Kopera, Applied Math, NPS Andreas Müller, Applied Math, NPS Simone Marras, Applied Math, NPS

### **Physical Parameterization and Databases**

Jim Doyle, NRL-Monterey Saša Gaberšek, NRL-Monterey Kevin Viner, NRL-Monterey Alex Reinecke, NRL-Monterey Eric Hendricks, NRL-Monterey

### **Time-Integrators and PETSc Interface**

Emil Constantinescu, ANL Debo Ghosh, ANL

### **Preconditioners and Iterative Solvers**

Carlos Borges, Applied Math, NPS Les Carr, Applied Math, NPS

### **Riemann Solvers and Limiters**

Dale Durran, University of Washington Maria Lukacova, University of Mainz

### **Many-Core Implementation**

Andreas Klöckner, Computer Science, UIUC Lucas Wilcox, Applied Math, NPS Tim Warburton, CAAM, Rice University Dave Norton, NVIDIA Daniel Abdi (soon to be) Applied Math, NPS

**ESMF Interface** Tim Campbell, NRL-Stennis Tim Whitcomb, NRL-Monterey

### **Data Structures Optimization**

Michael Bader, Computer Science, TUM Kaveh Rahnema, Computer Science, TUM Alex Breuer, Computer Science, TUM



motivation

results

Goal

# methods

- dynamic AMR, uniform meshes
- high order, low order

motivation

results

# applications

cloud simulations

next steps

Hurricane simulations



Goal

# methods

- dynamic AMR, uniform meshes
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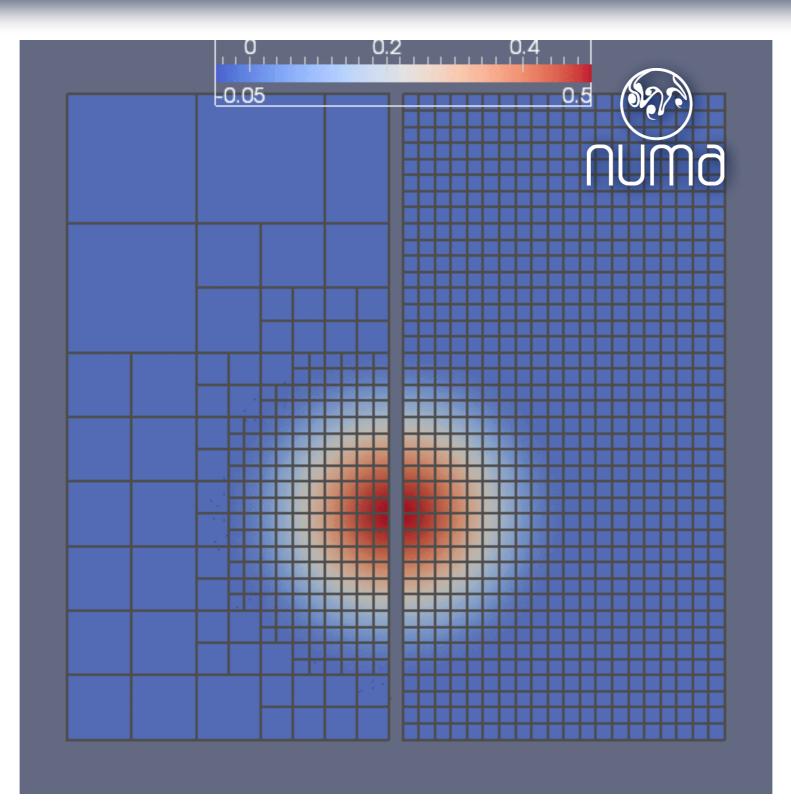
Hurricane simulations

For which of these applications should we use these methods and how should we use them?

results



### **Motivation** Warm air bubble test case with $\mu = 0.1 \text{m}^2/\text{s}$

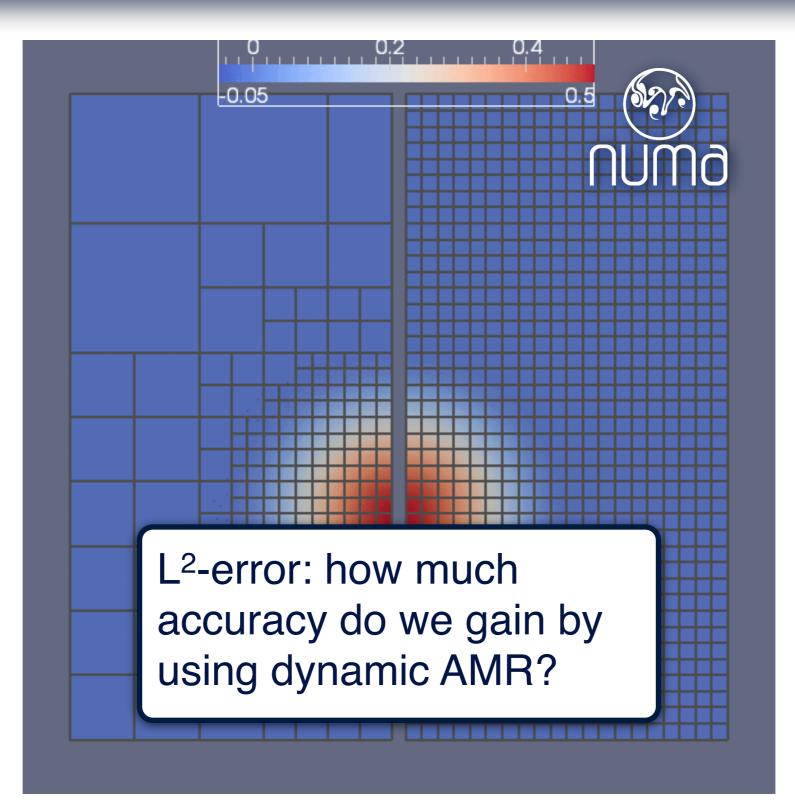




motivation

results

### **Motivation** Warm air bubble test case with $\mu = 0.1 \text{ m}^2/\text{s}$





motivation

results

### **Questions for Today** Warm air bubble test case with $\mu = 0.1 \text{ m}^2/\text{s}$ at t = 700 s

### **Questions**:

L<sup>2</sup>-error: how much accuracy do we gain by using dynamic AMR?



**Results**:



motivation

results

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- 2. How does the benefit of AMR depend on the initial condition?
- 3. What is the benefit for the error of  $max(\theta)$ ?



**Results**:

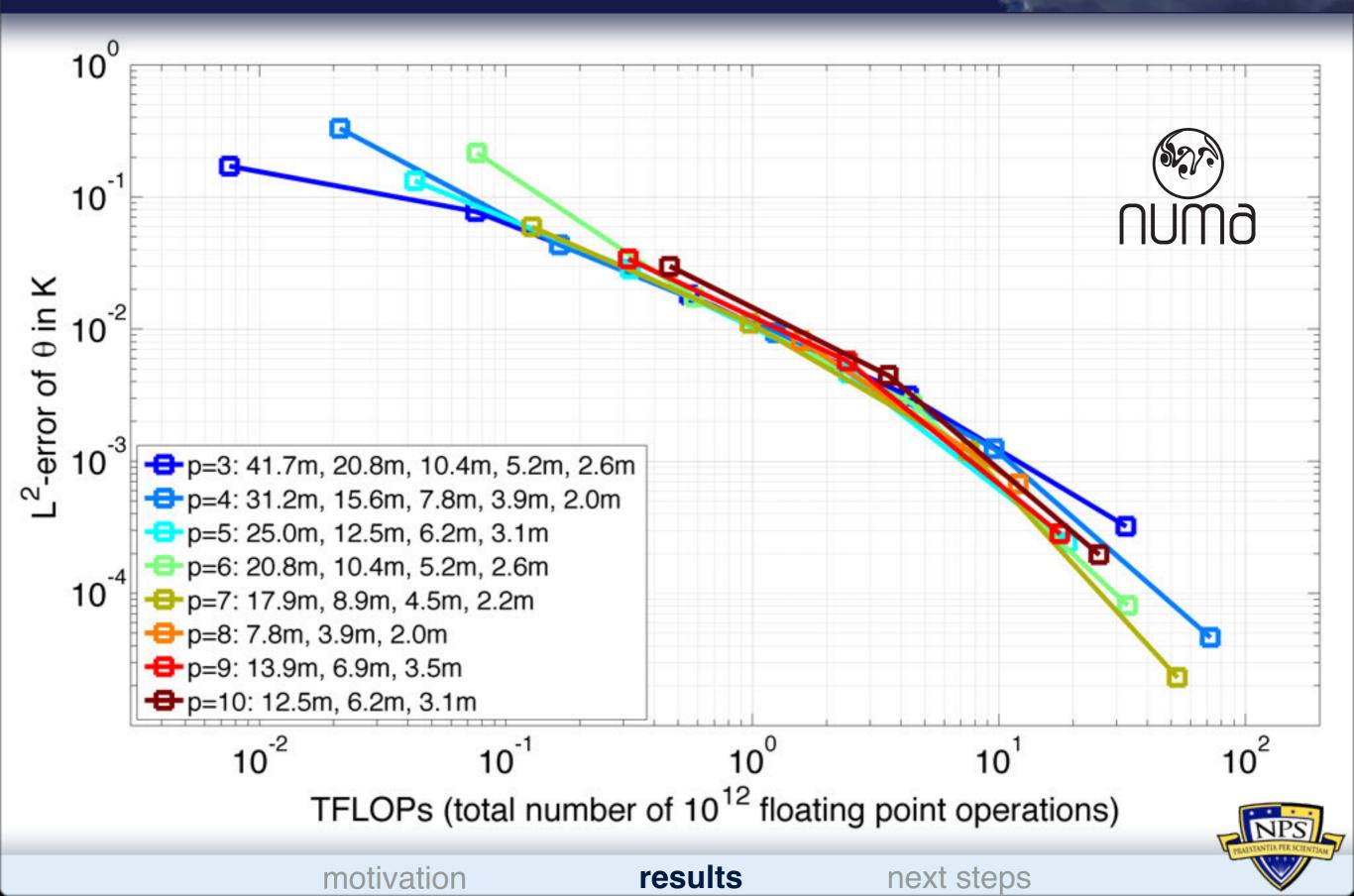
next steps

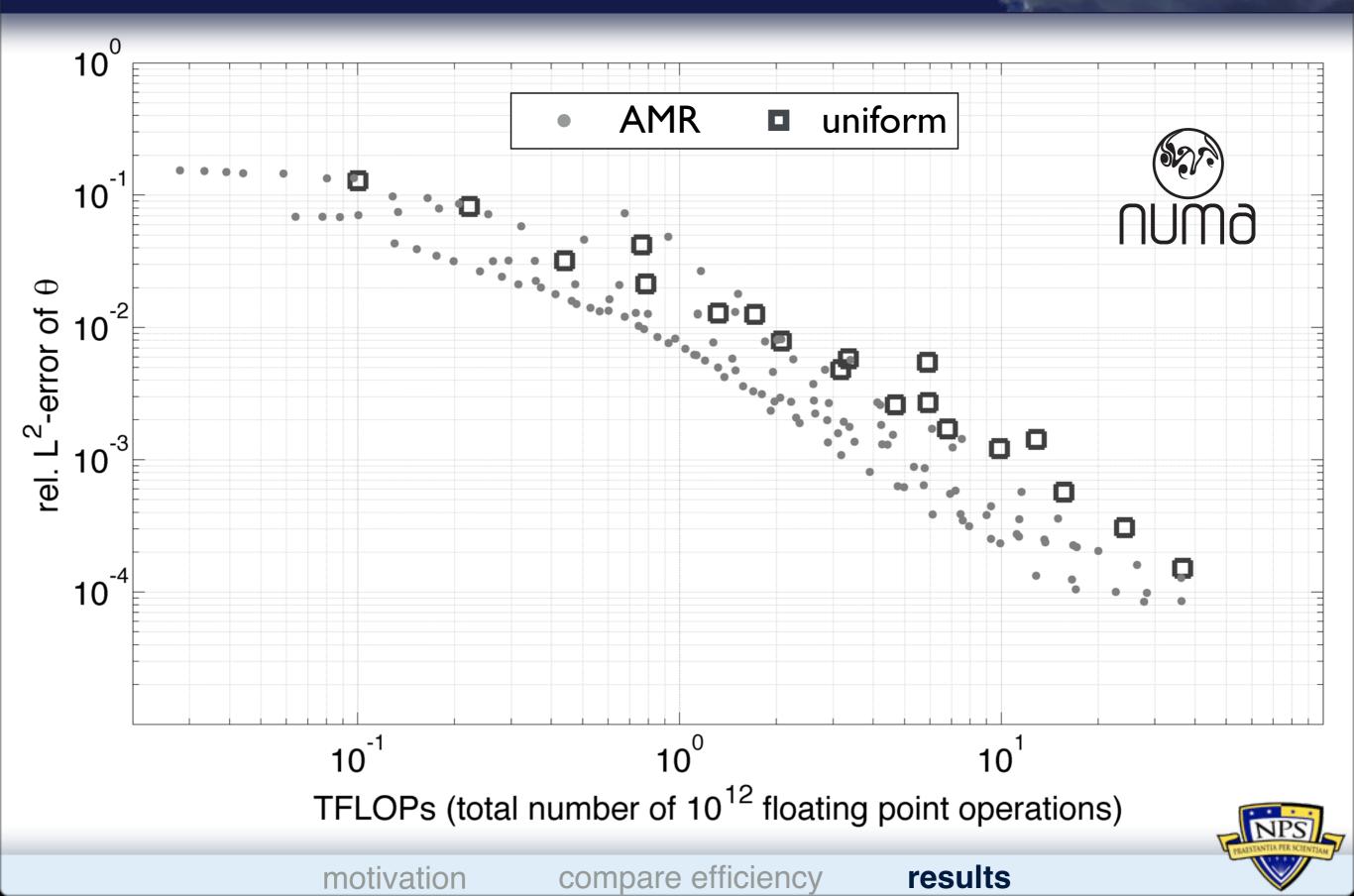


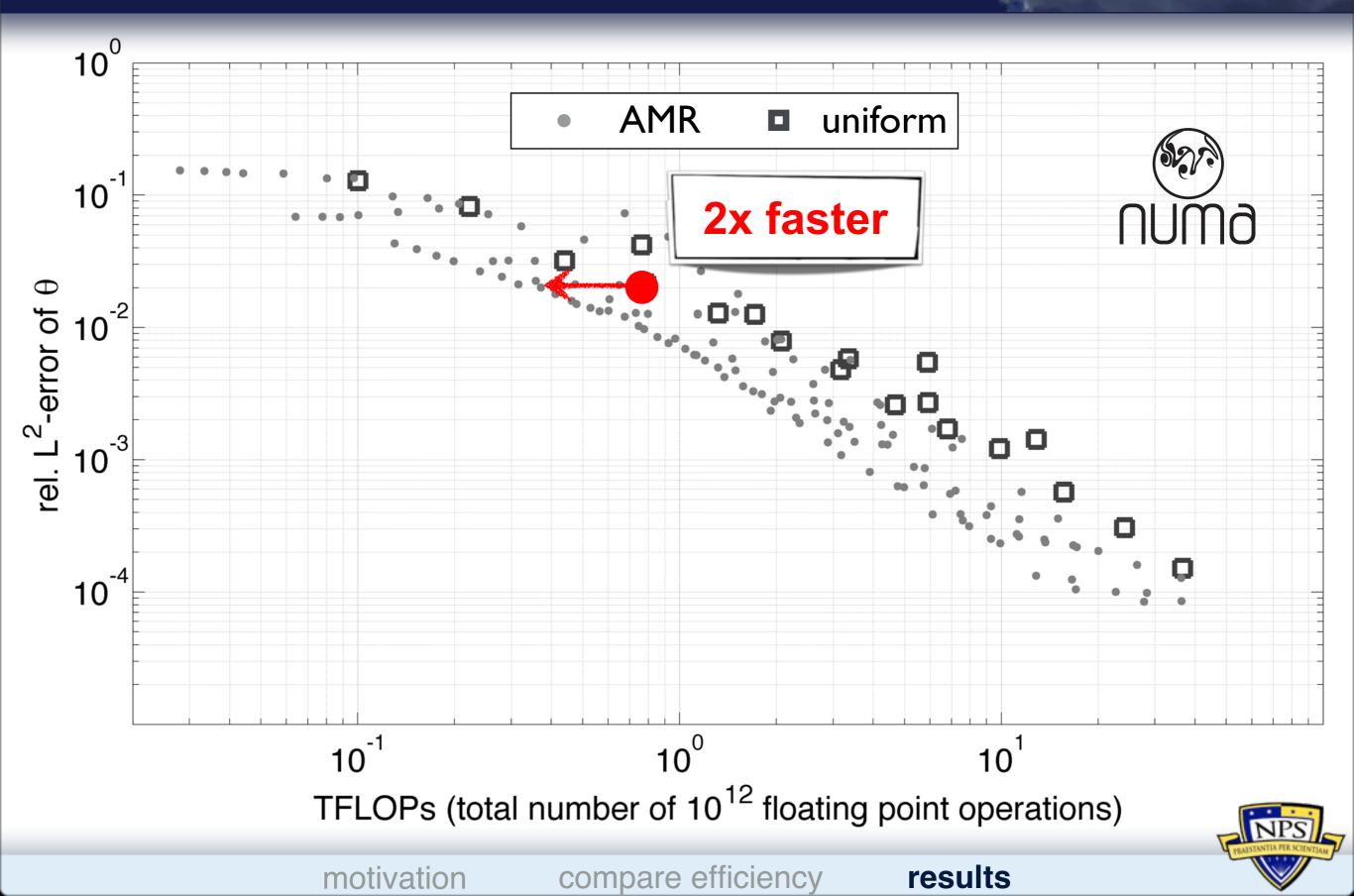
motivation

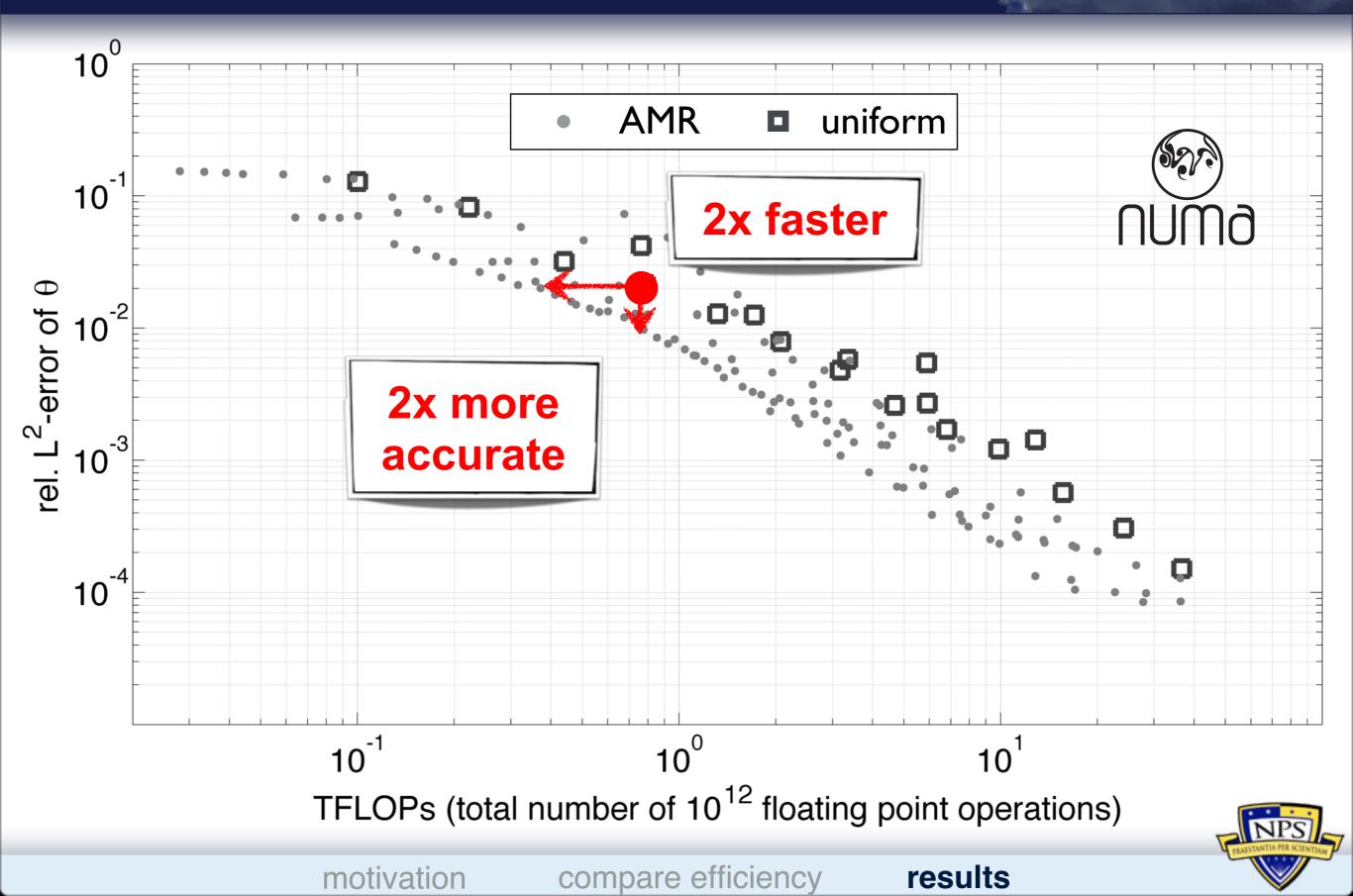
results

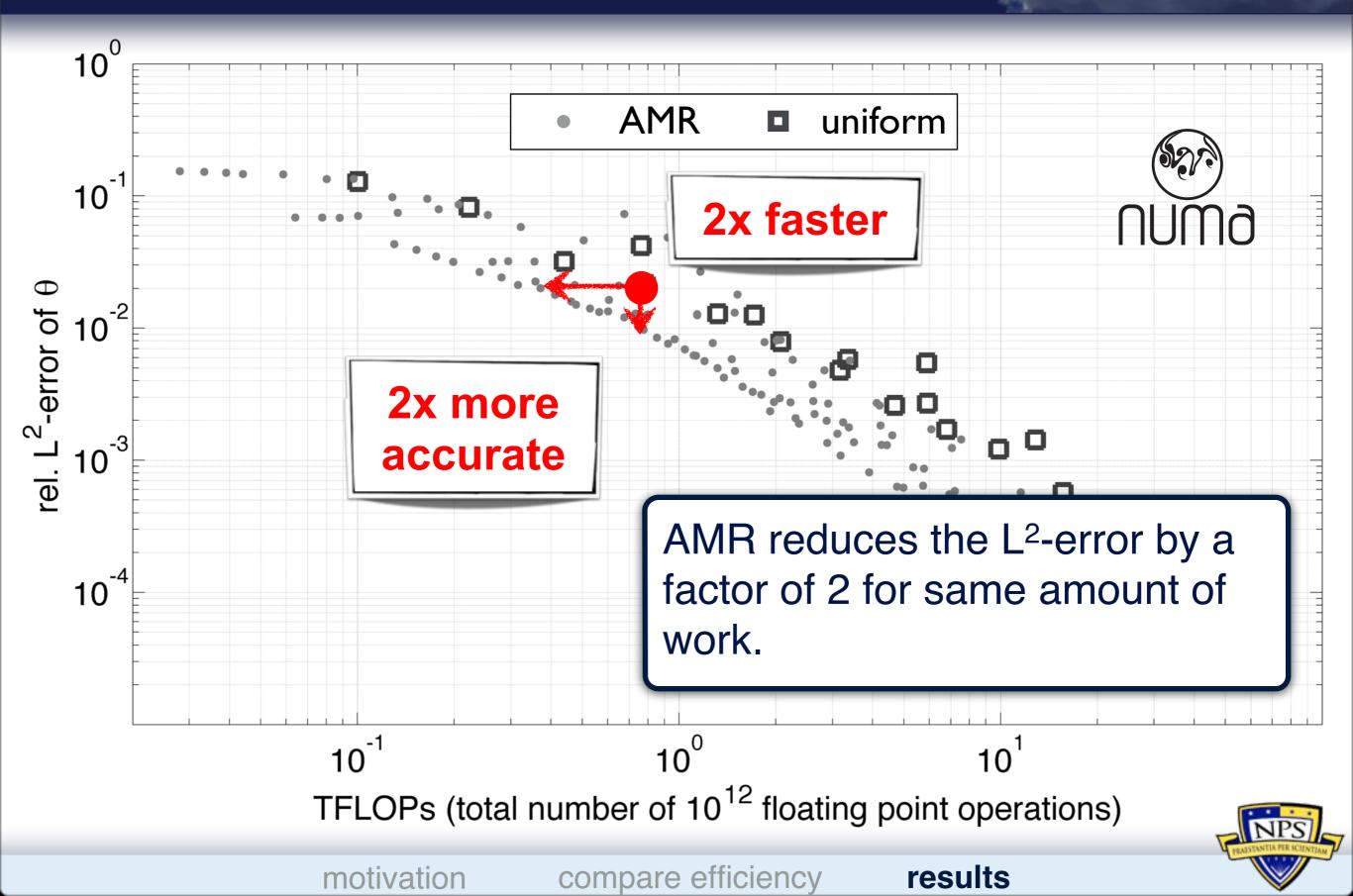
## L<sup>2</sup>-error of uniform simulations











### **Questions**:

1. L<sup>2</sup>-error: how much accuracy do we gain by using dynamic AMR?

### **Results**:

AMR reduces the L<sup>2</sup>-error by a factor of 2 for same amount of work.

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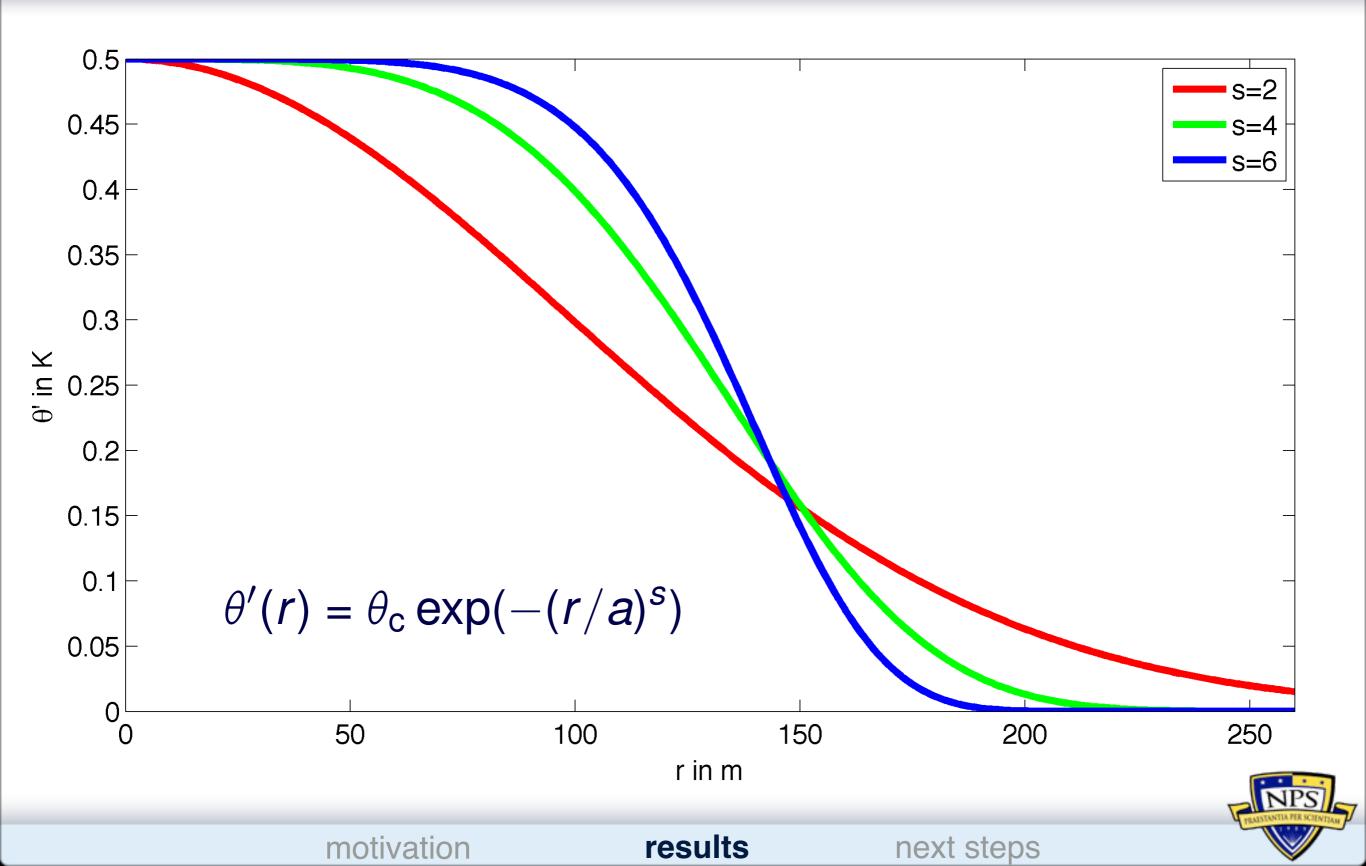
motivation



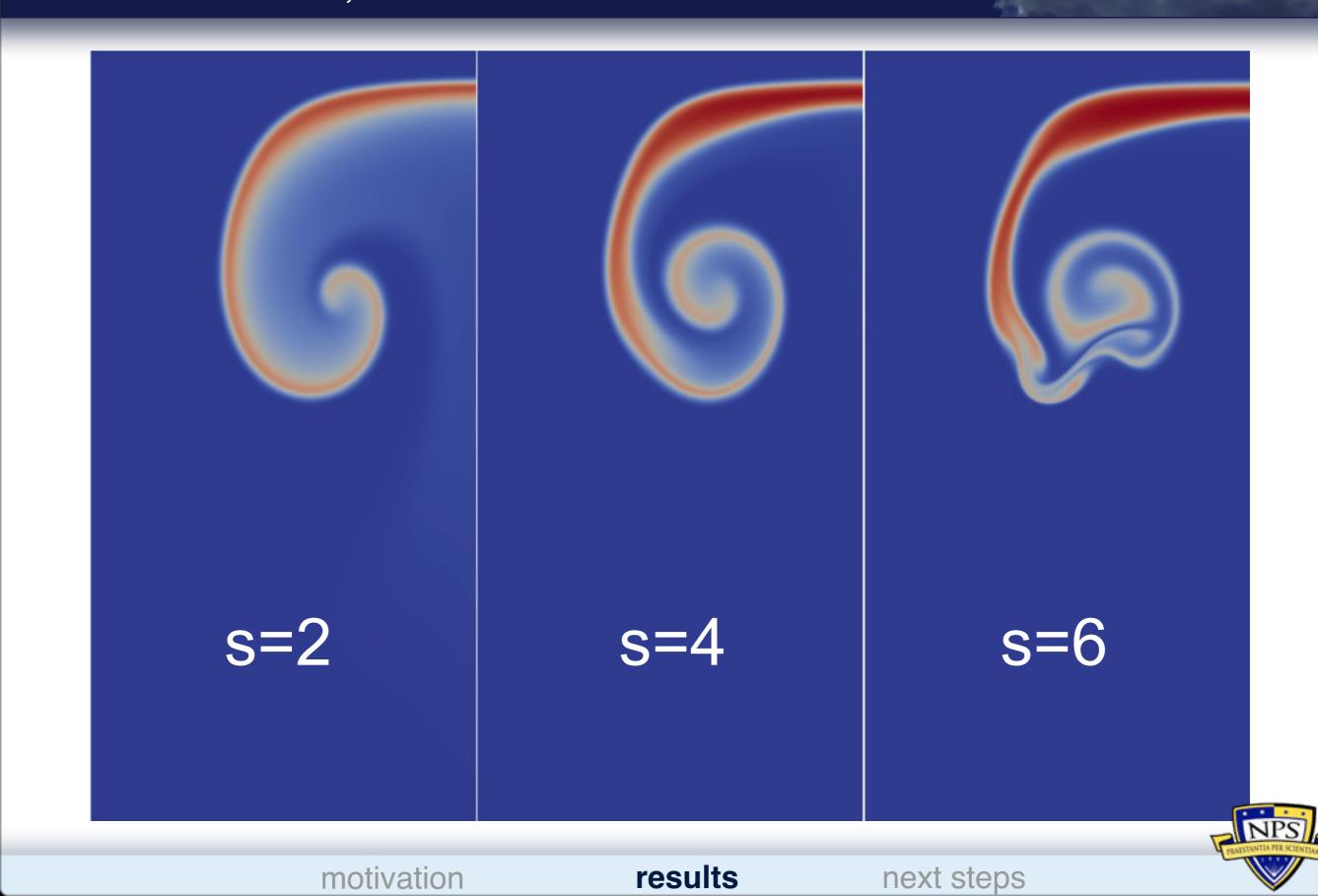
results

## Three different initial profiles for $\theta'$

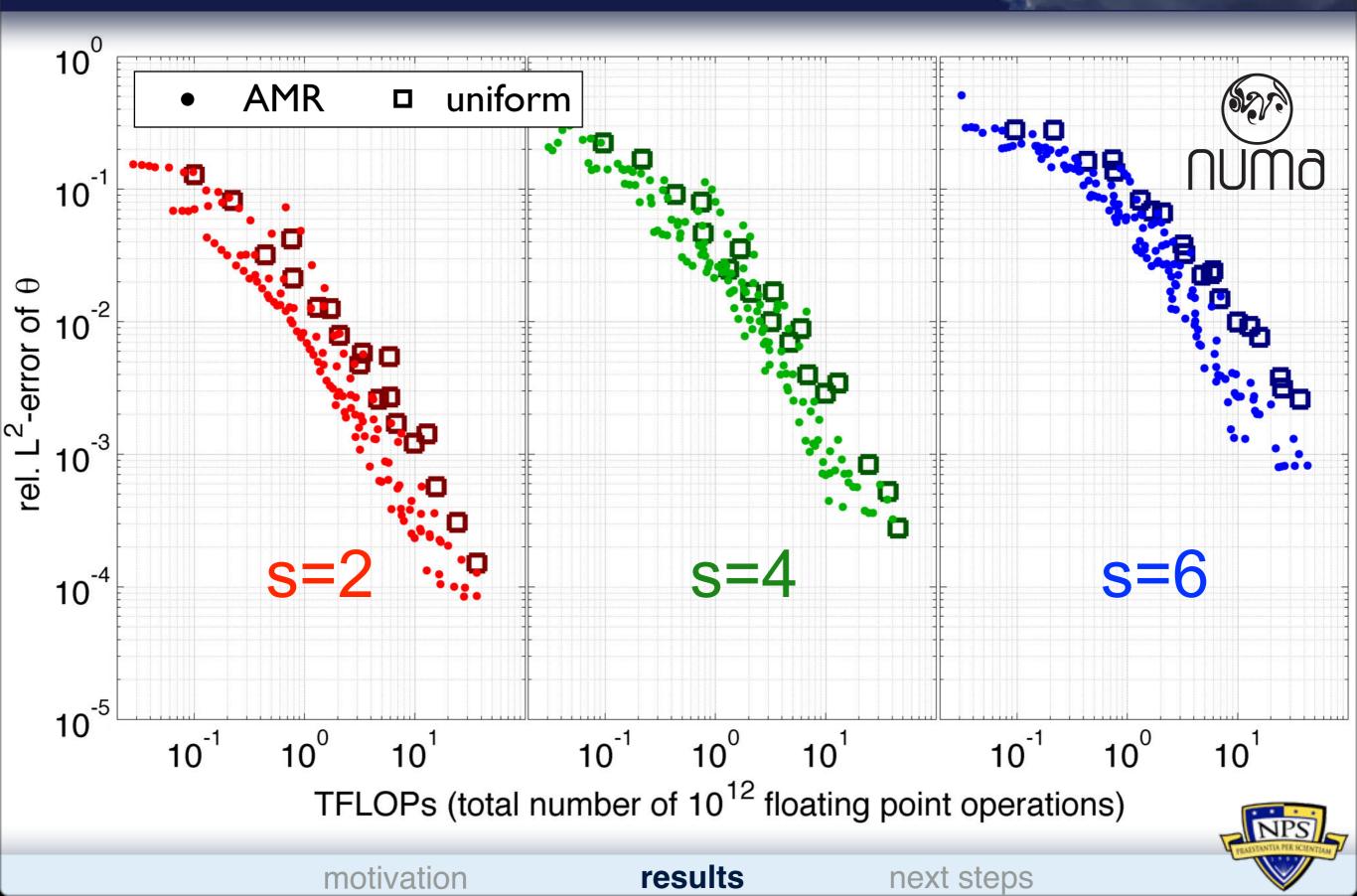
as a function of distance from the center of the bubble r



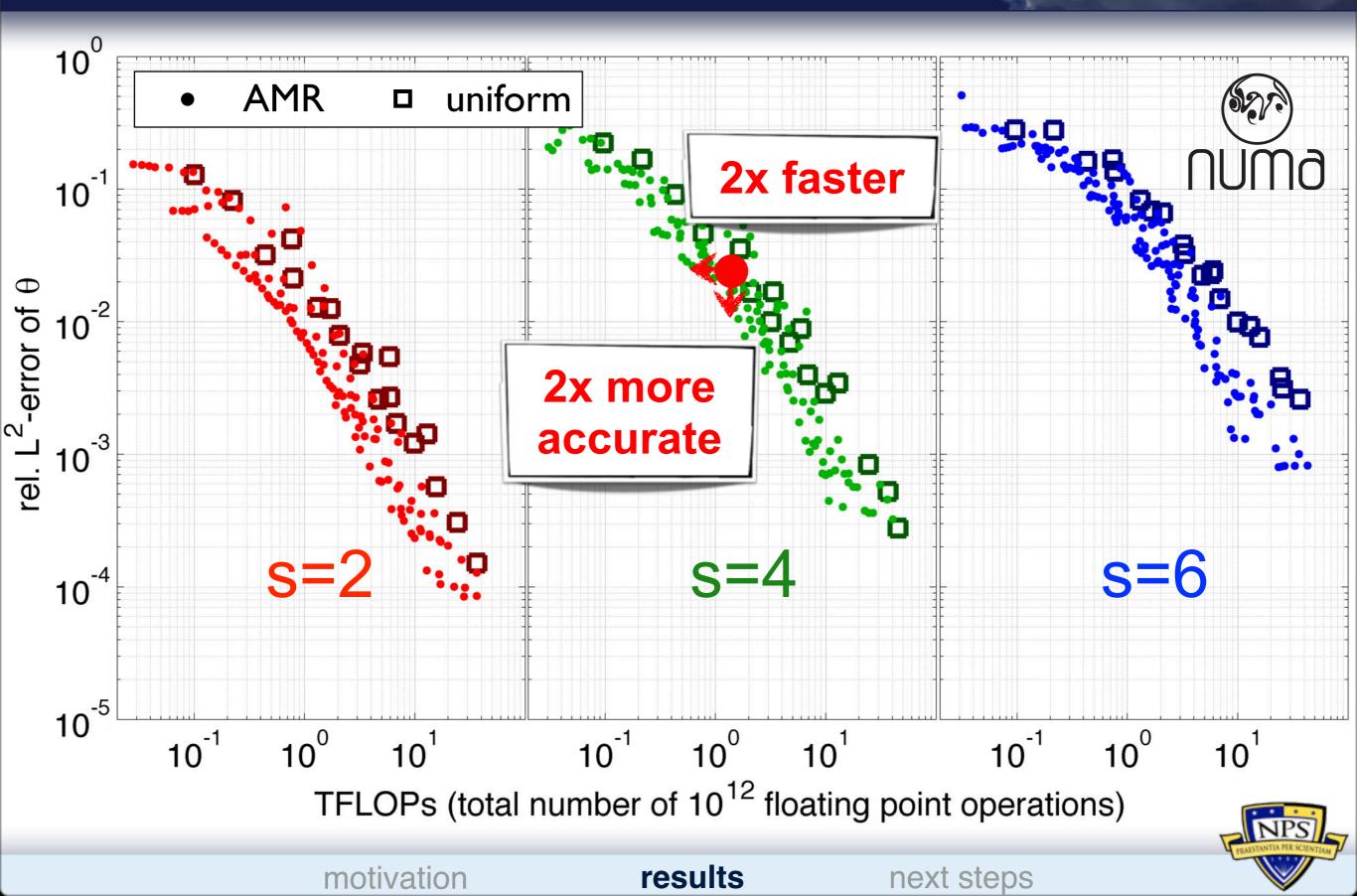
### **Reference result for modified bubbles** resolution $\Delta x = 40$ cm, time t=700s



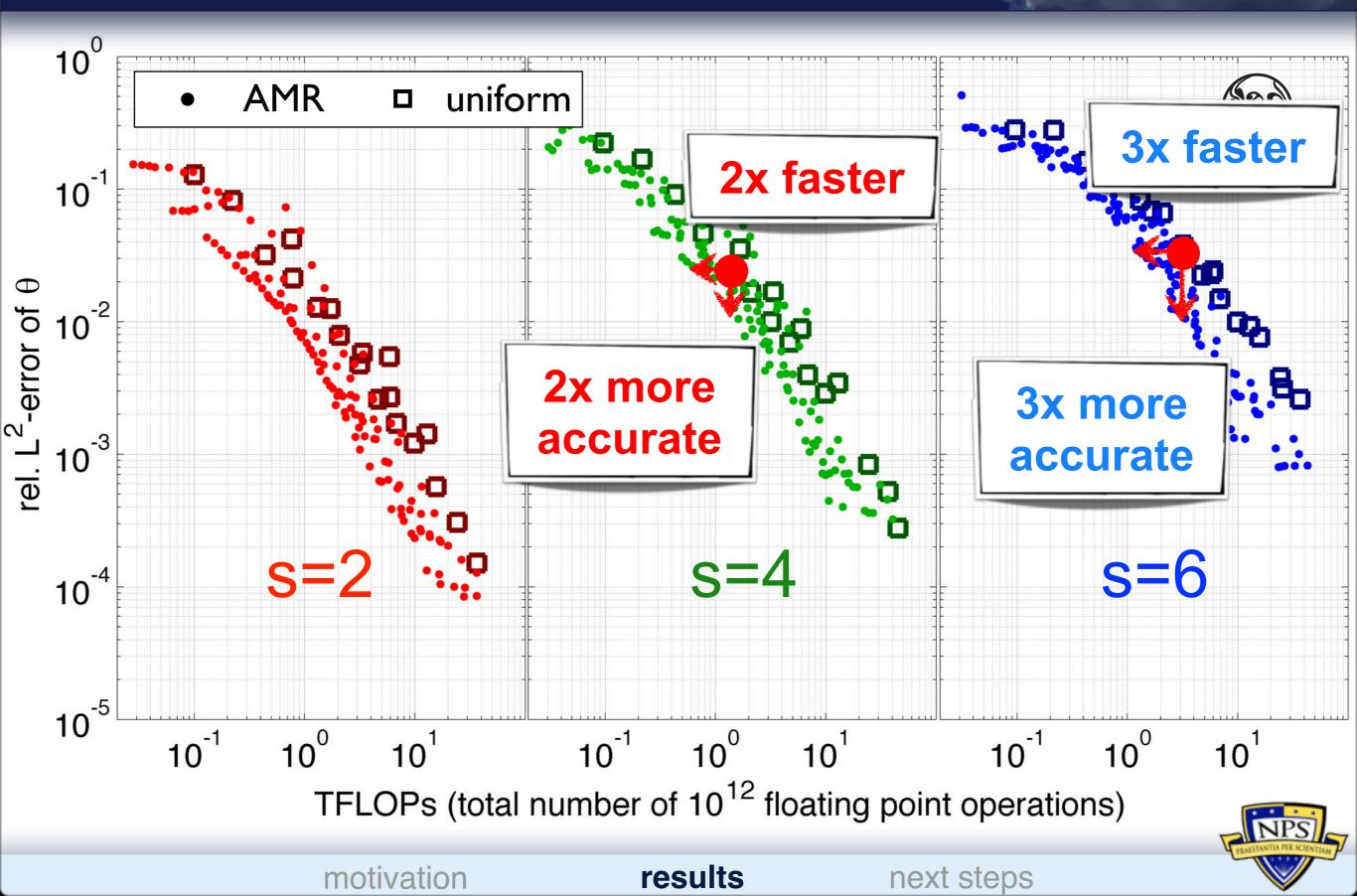
### L<sup>2</sup>-error for 3 different initial conditions viscosity: $\mu$ =0.1m<sup>2</sup>/s



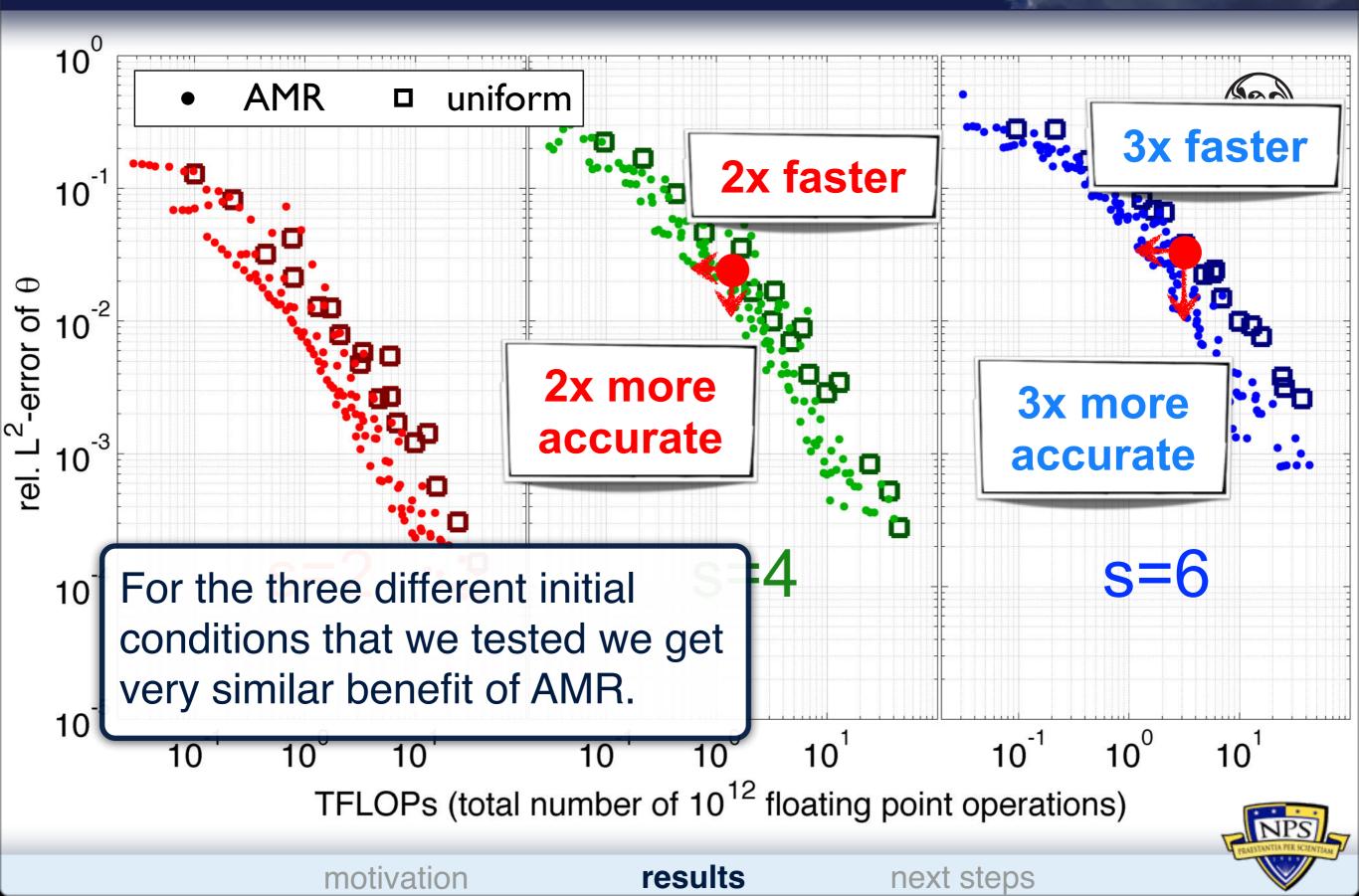
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next steps

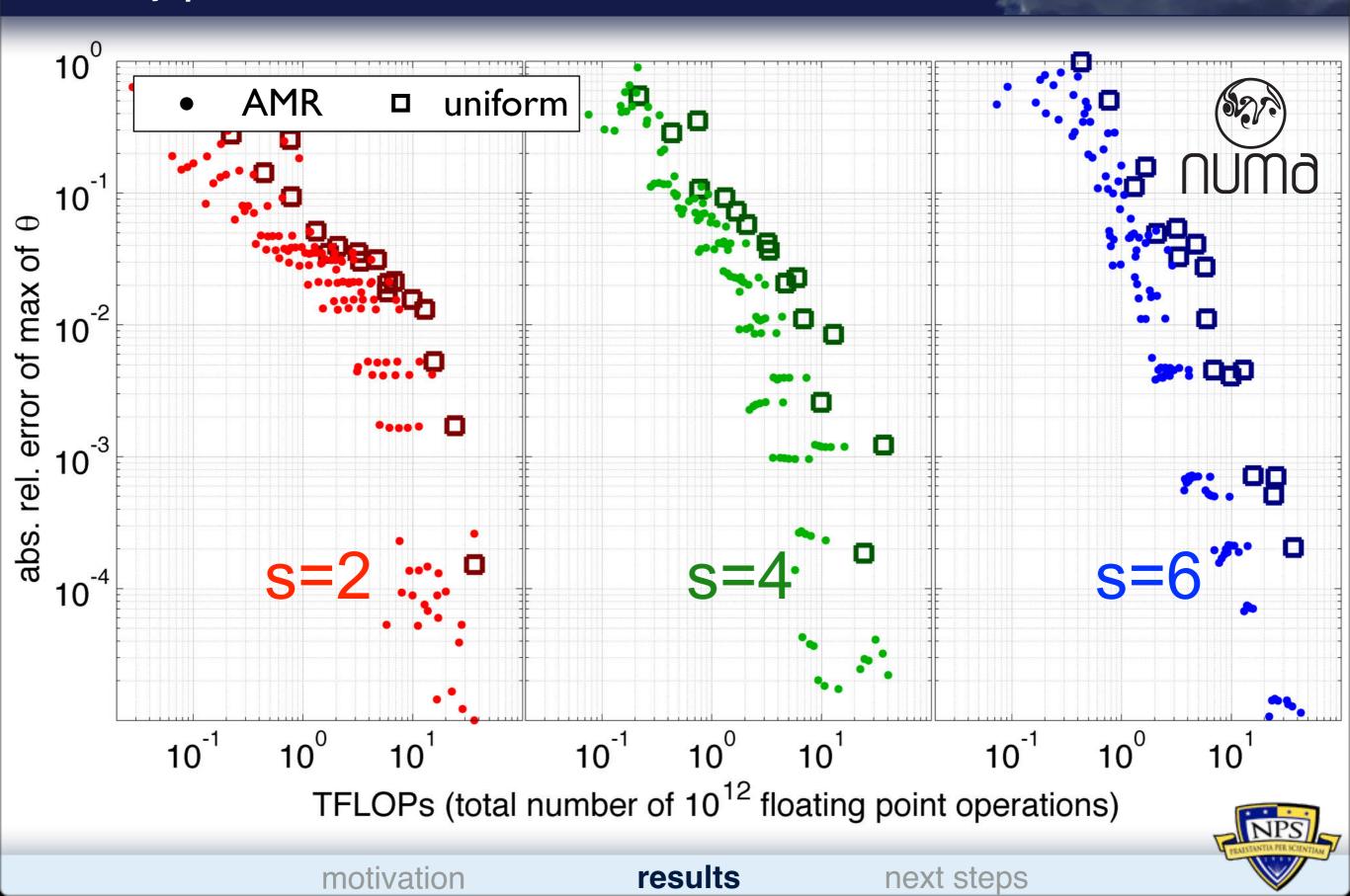
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motivation

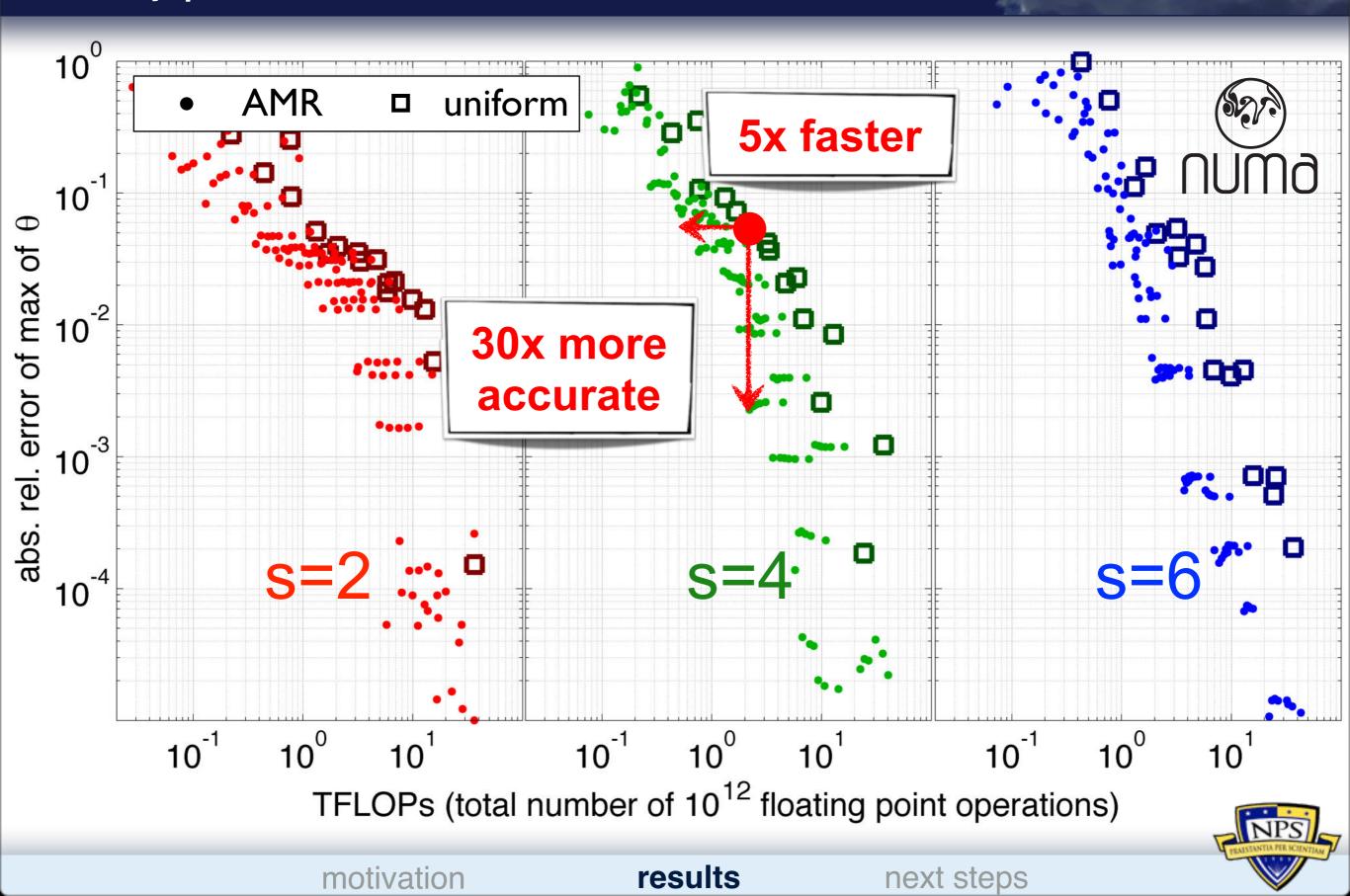
results



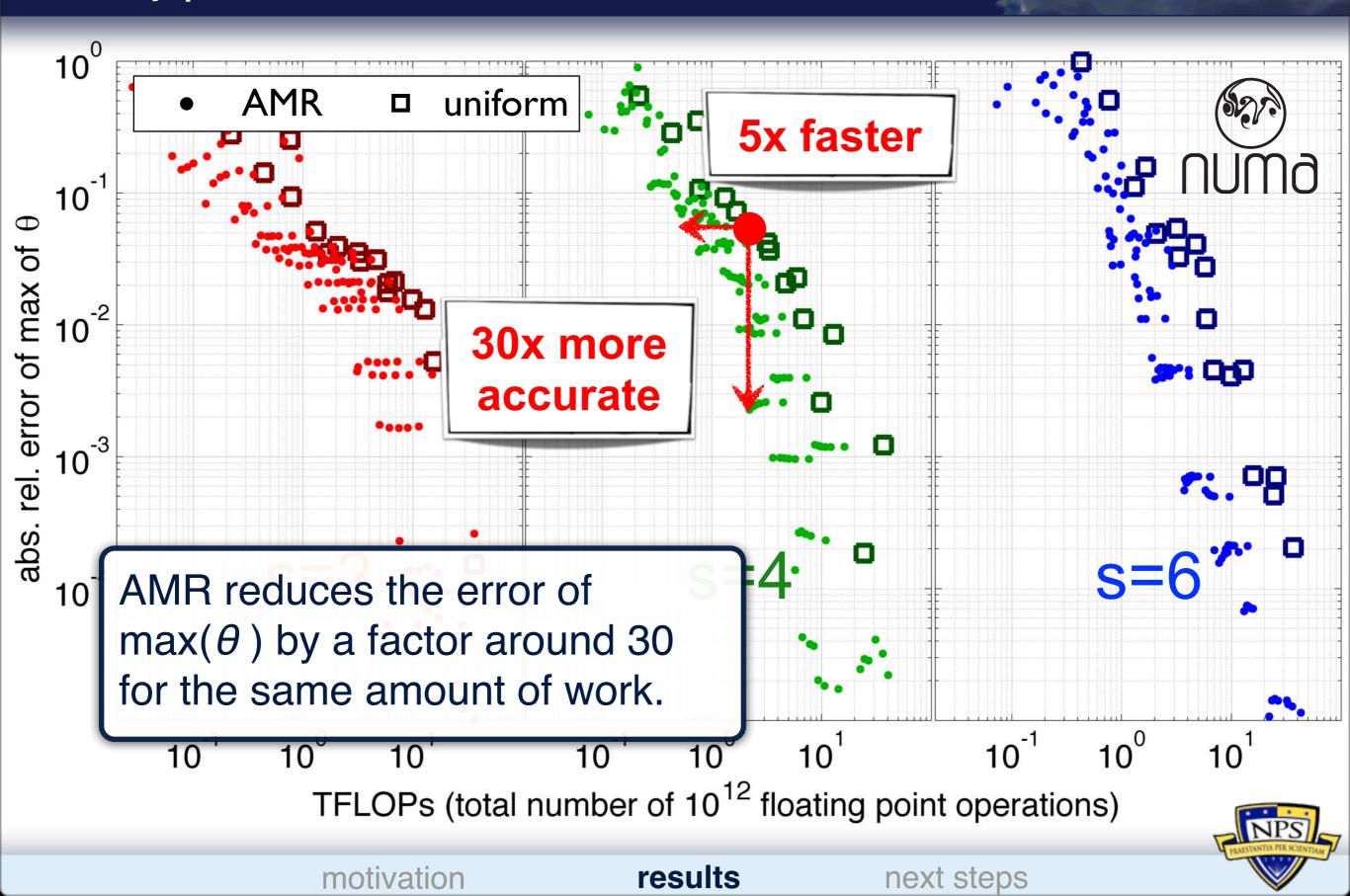
### absolute relative error of $max(\theta)$ viscosity: $\mu=0.1m^2/s$



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motivation

results

AMR reduces the error of  $max(\theta)$  by a factor around 30 for the same amount of work.





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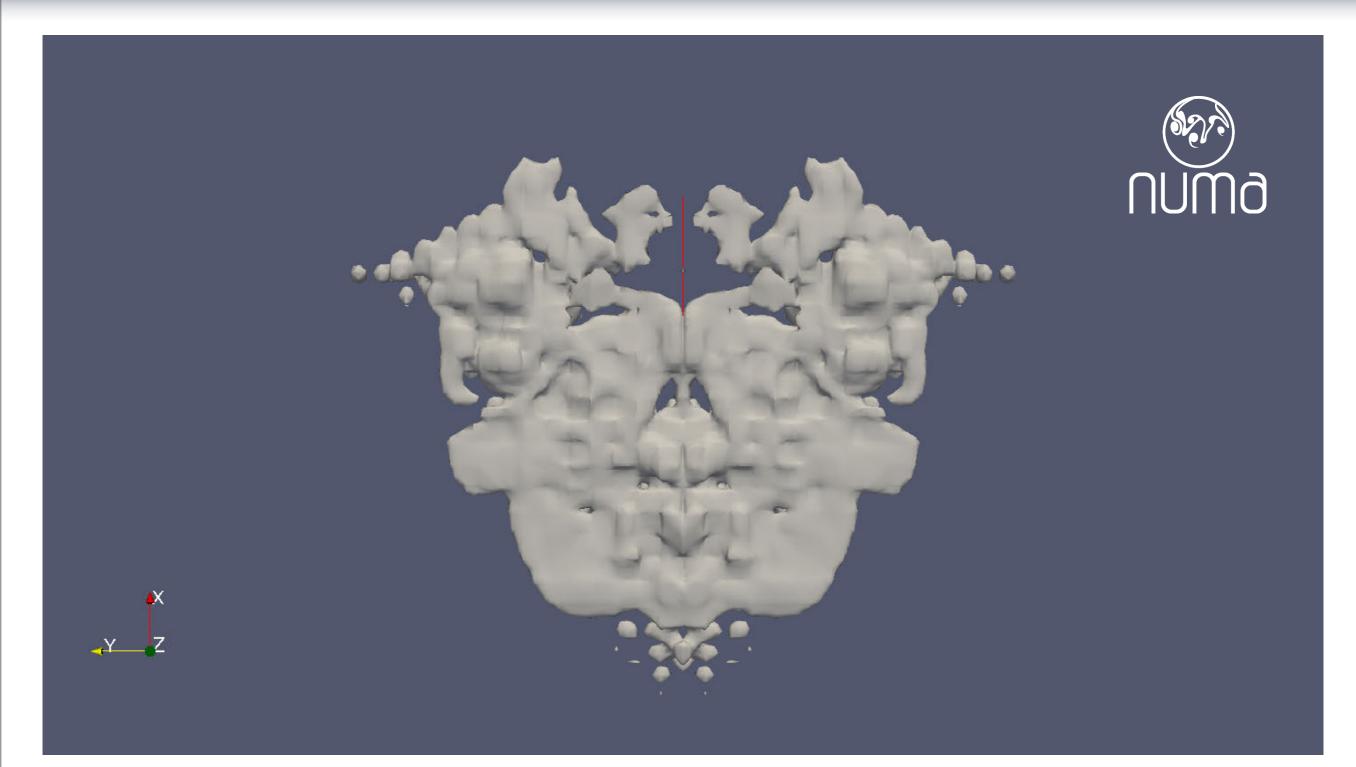
next steps

Next steps: 1. different refinement criteria (gradient, ...), 2. include moisture, 3. run comparison in 3D

results



# Squall line simulation with NUMA isosurface of cloud water content $q_c=0.0035$ at t=7500s

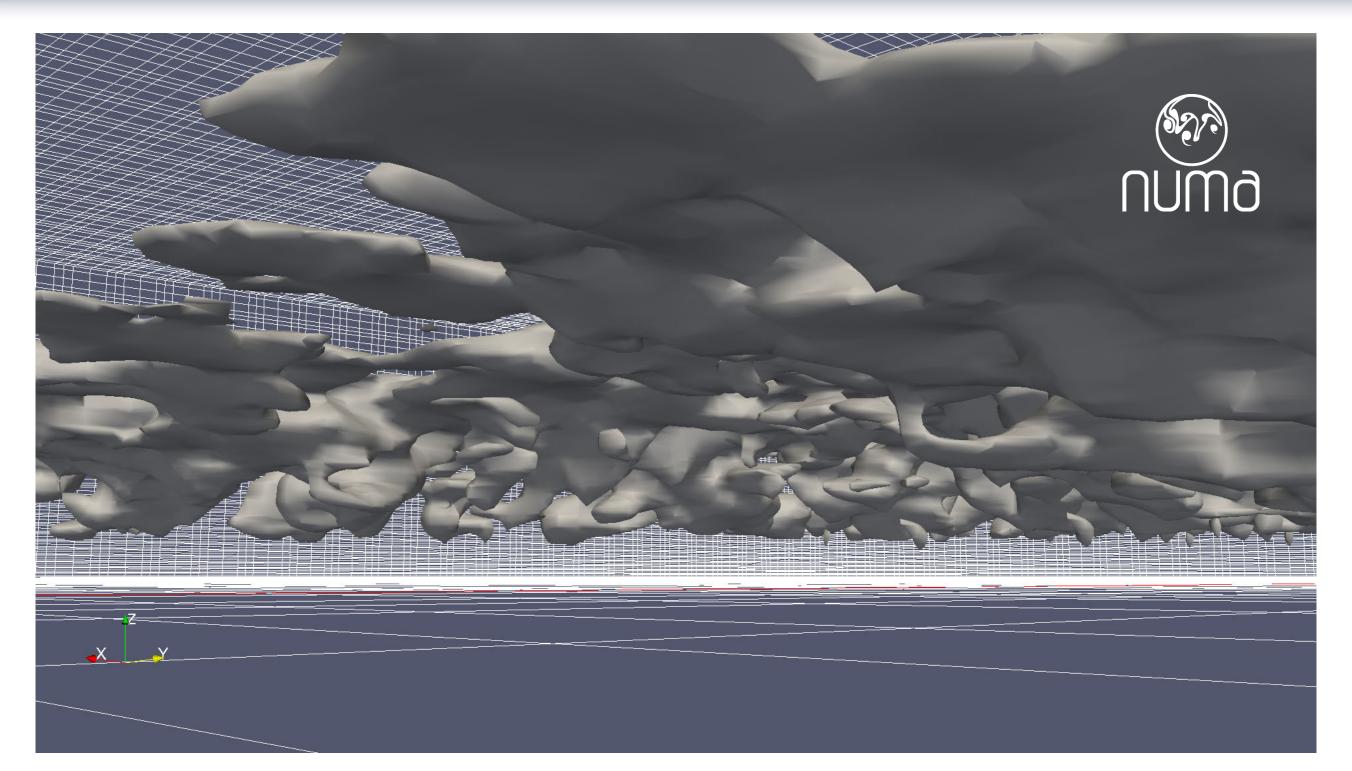




motivation

results

# Squall line simulation with NUMA isosurface of cloud water content $q_c=0.0035$ at t=7500s





results

motivation

# Squall line simulation with NUMA visualization with Maya® (see <a href="http://anmr.de">http://anmr.de</a> for instructions)





motivation

results



## Thank you for your attention!