

# Fast Algorithms for Particle-Based Simulations

**Stan Moore**

**April 3, 2014**

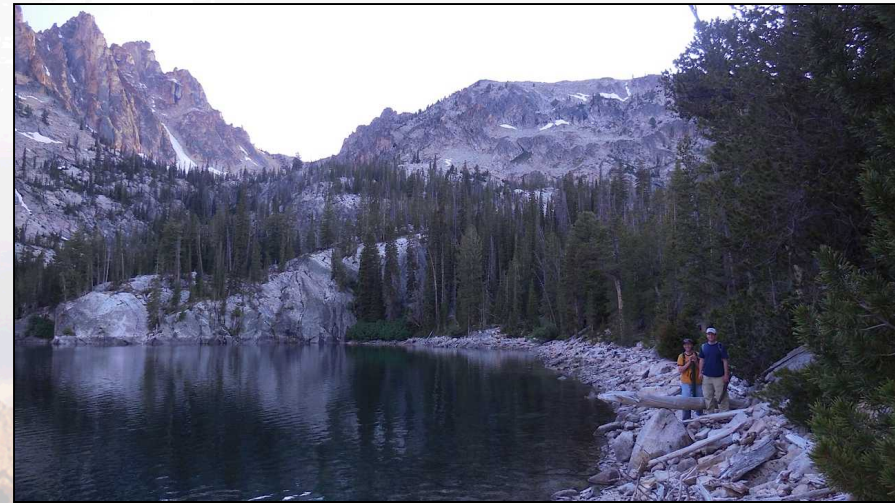
**1444 Interview Candidate**

Sandia National Laboratories is a multi-program laboratory managed and operated by Sandia Corporation, a wholly owned subsidiary of Lockheed Martin Corporation, for the U.S. Department of Energy's National Nuclear Security Administration under contract DE-AC04-94AL85000.



# A little about me ...

- Grew up on a farm in Filer, Idaho
- Love outdoor activities (backpacking, hiking, fly fishing, etc.)
- BS Chemical Engineering, Brigham Young University, 2008
- PhD Chemical Engineering, Brigham Young University, 2012
- Dissertation topic: Developing a new method to predict chemical potential using molecular simulations





# Sandia Projects

- Been at Sandia since August, 2012

## Projects:

- **LAMMPS (molecular dynamics) code development:** Added and improved long-range electrostatic methods
- **Modeling of a dipolar fluid in an external electric field** (uses LAMMPS)
- **Advanced force field development (SNAP)** (uses LAMMPS, working to develop a quantum accurate  $\text{SiO}_2$  force field)
- **Aleph (PIC-DSMC) code development:** Added an automatic mesh coarsening method for particle interactions (based on oct-tree algorithm)
- **Modeling of triggered vacuum gap switches**

Thanks to Paul Crozier for being a great mentor during my time at Sandia







# Importance of Algorithms

---

- Computing resources limited
- Gains in processor speed not as dramatic as before
- Increasing number of processors increases power consumption
- Better algorithms can save time and money
- Example: Fast Fourier transform (FFT) vs traditional discrete Fourier transform





# Introduction

- Hybrid particle-in-cell (PIC) and direct simulation Monte Carlo (DSMC) methods are frequently used to simulate low density interacting plasmas
- A single mesh is often used for both PIC and DSMC calculations
- The mesh size for PIC is often limited by the Debye length
- The collision cell size for DSMC is limited by the mean free path (can be much larger than the Debye length)
- Too few computational particles per DSMC collision cell can lead to errors
- Therefore, the optimal PIC mesh may be suboptimal for calculating DSMC collisions



Parts of this work were done in collaboration with Paul Crozier, Chris Moore, and Matt Bettencourt



Sandia National Laboratories



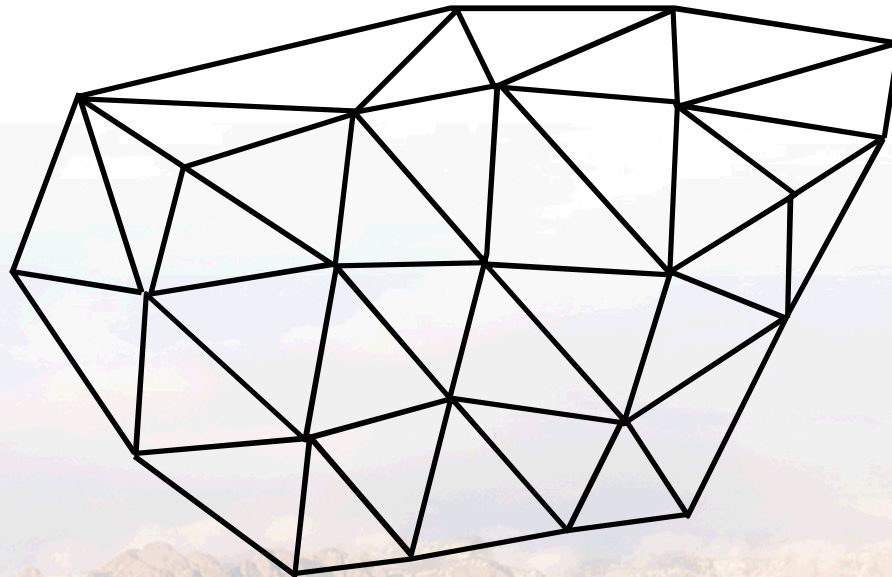
# Overview of New Patching Method

- Use a fine mesh for PIC (unstructured)
- Use a rectangular grid to conglomerate many PIC elements into a single DSMC collision cell
- Size DSMC collision cells based on mean free path,  $\lambda_{\text{mfp}}$
- Use oct-tree algorithm to adjust the size of DSMC collision cells on the fly



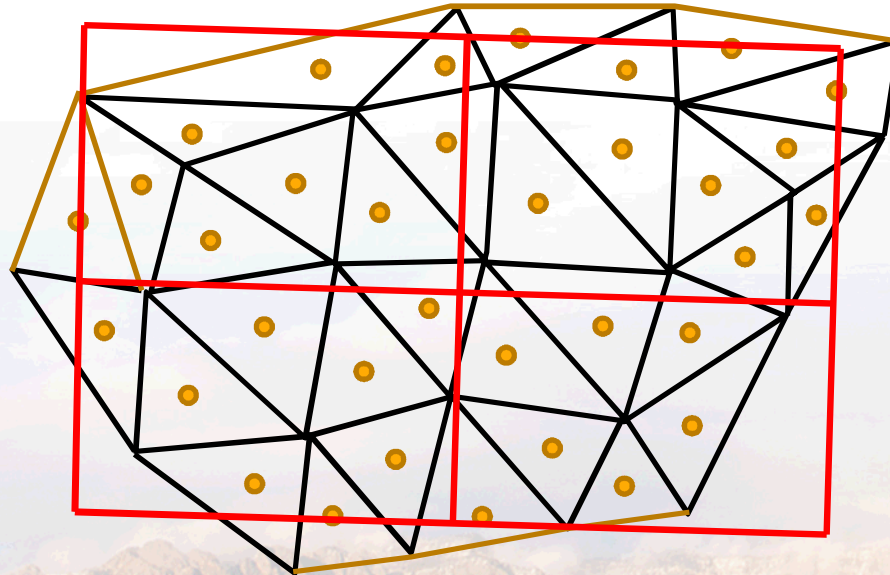
# Patch Method

- Original (unstructured) PIC mesh



# Patch Method

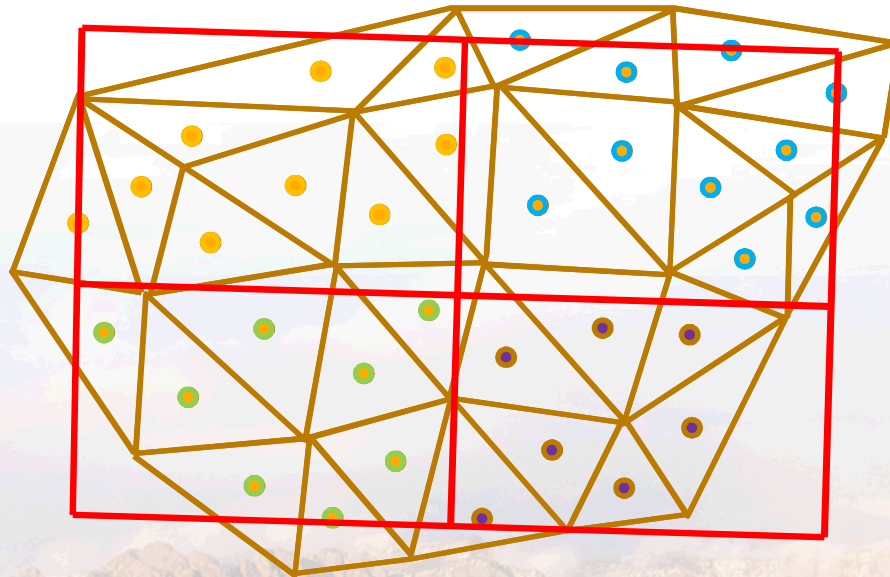
- Apply rectilinear grid based on element centroid



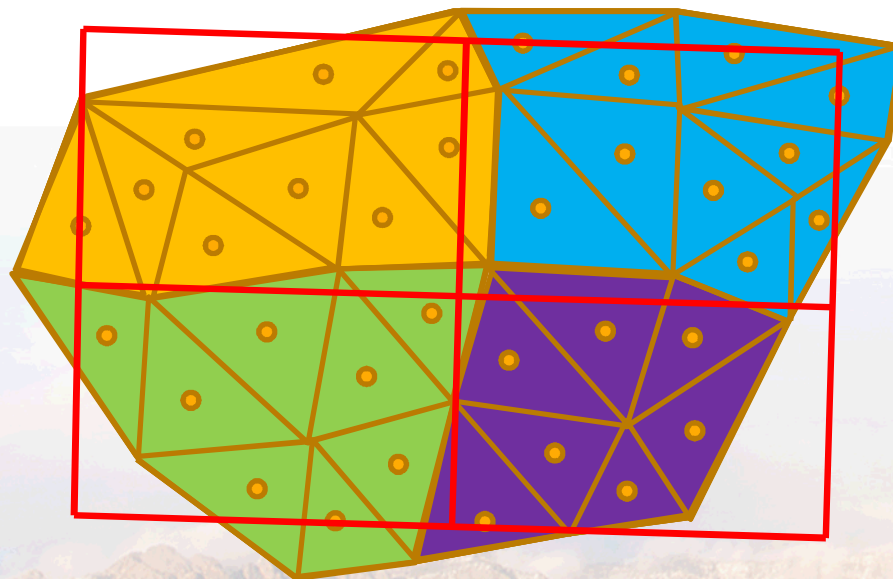


# Patch Method

- Assign elements to patches (based on element centroid)

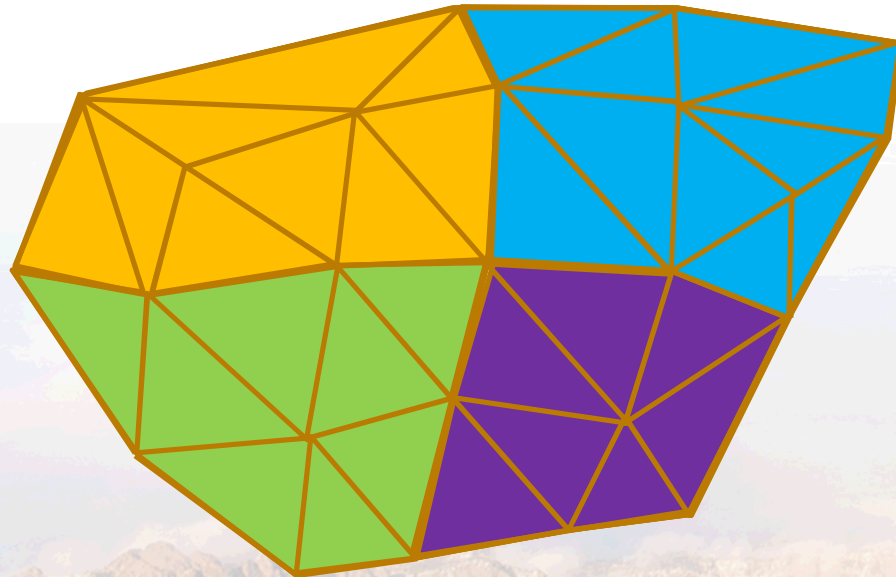


# Patch Method



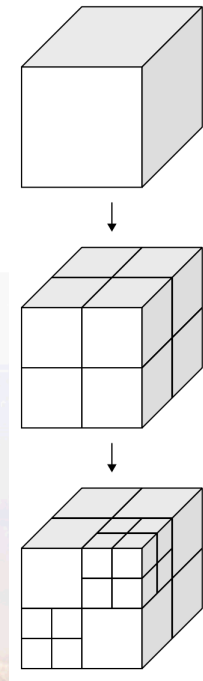
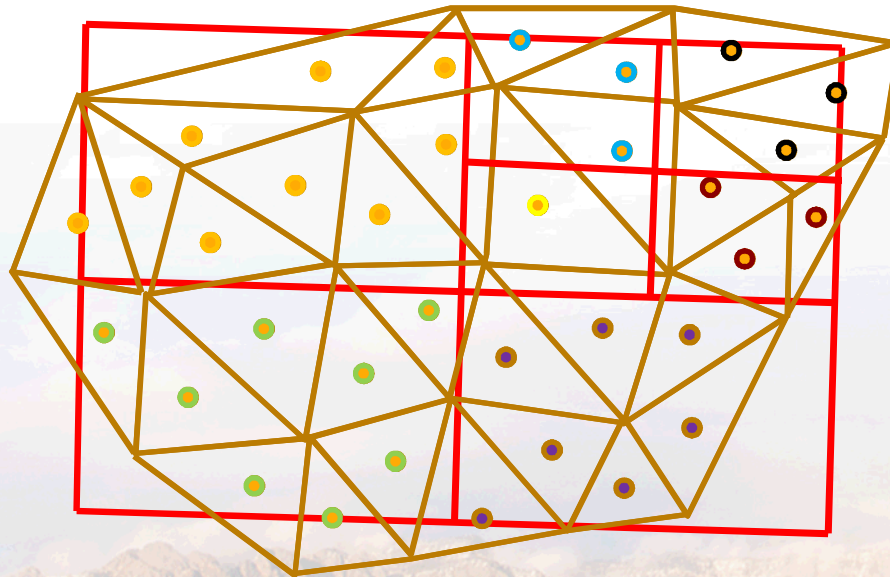
# Patch Method

- Patched mesh
- Use patches to compute DSMC collisions



# Oct-tree Refinement

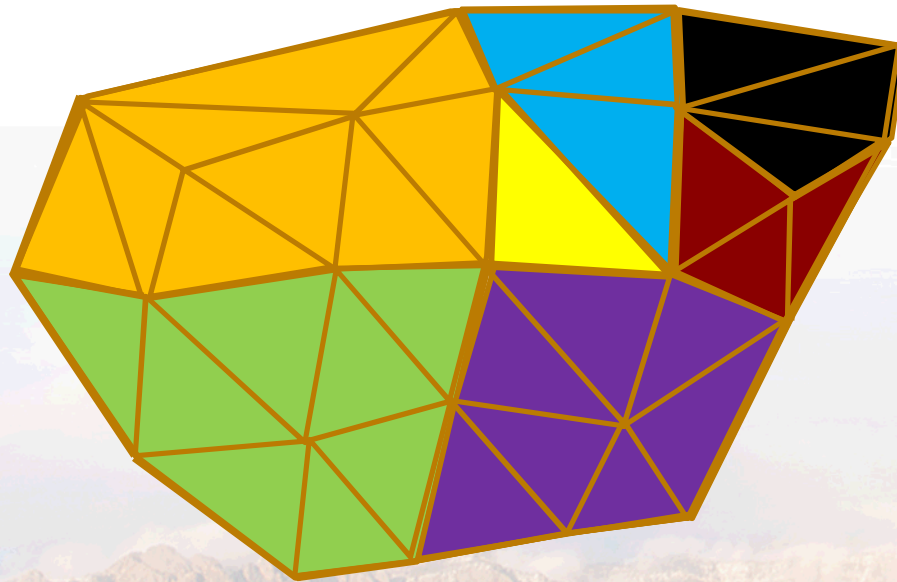
- Use oct-tree algorithm to refine mesh based on mean free path





# Oct-tree Refinement

- Refined mesh



# Temporal Averaging

- Calculate  $\lambda_{\text{mfp}}$  for each element as:

$$\lambda_{\text{mfp}} = \frac{v}{Z} n$$

$v$  = velocity  
 $Z$  = interaction frequency  
 $n$  = number of particles

- With a high computational particle weighting, temporal smoothing is needed. Can use either:

$$\langle \lambda_{\text{mfp}} \rangle = \left\langle \frac{v}{Z} n \right\rangle \qquad \langle \lambda_{\text{mfp}} \rangle = \frac{\langle v \rangle}{\langle Z \rangle} \langle n \rangle$$

- Sometimes can get zero interactions in a timestep. With the first option, this leads to division by zero
- Found that the second option works much better





# Automatic Sizing of Patches

Patch size is dynamically adjusted based on the local mean free path  $\lambda_{\text{mfp}}$ :

1. Compute  $\lambda_{\text{mfp}}$  for each interaction on an elemental basis (using all species)
2. For each interaction, average  $\lambda_{\text{mfp}}$  over elements in the oct-tree cell
3. Take the minimum of all the average  $\lambda_{\text{mfp}}$  and divide by 2, use this to size patches using the oct-tree algorithm





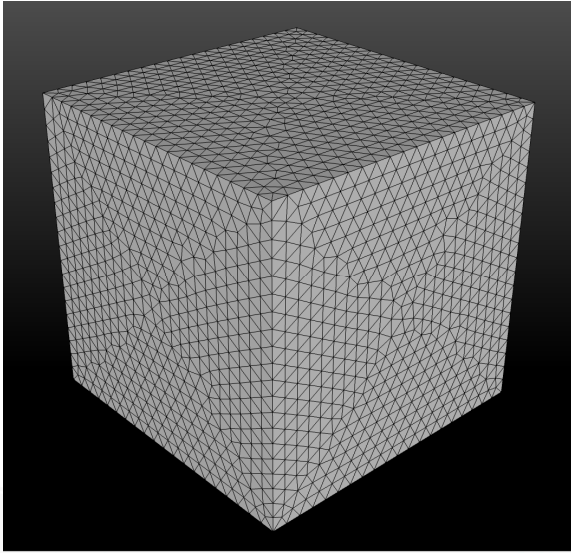
# HI Test Problem

- Hydrogen iodide (HI) molecules interacting with elastic collisions
- Use variable hard sphere (VHS) interaction cross section
- $T = 594.6 \text{ K}$ ,  $n = 10^{20} \text{ m}^{-3}$
- 3D cubic system,  $L = 5 \text{ cm}$
- Analytic  $\lambda_{\text{mfp}} = 3.67 \text{ cm}$
- $0.5 * \lambda_{\text{mfp}} \Rightarrow$  minimum of 27 patches, but oct-tree on cube uses powers of 8  $\Rightarrow$  64 patches

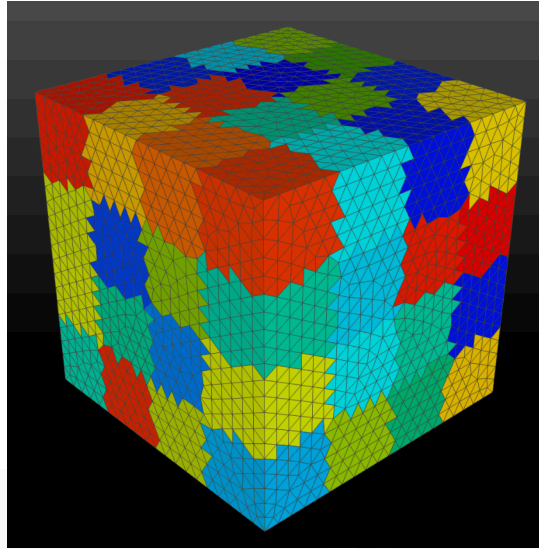




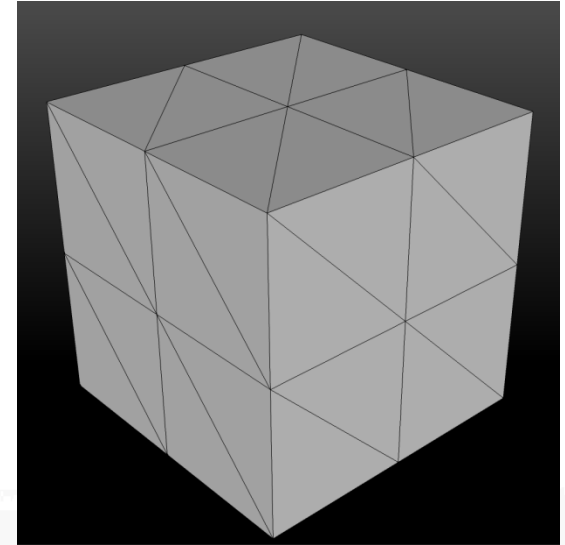
## 3 Meshes



*fine* mesh  
56,557 elements  
1,696,710 particles



*patched* mesh  
64 patches  
1,920 particles



*coarse* mesh  
96 elements  
2,880 particles

- Adjust particle weighting → approximately 30 particles per element or patch





# Accuracy Comparison

Ran each simulation for 3 hours ( $\Delta t = 10^{-5}$  s) on one processor, repeated 6 times with a different random number seed

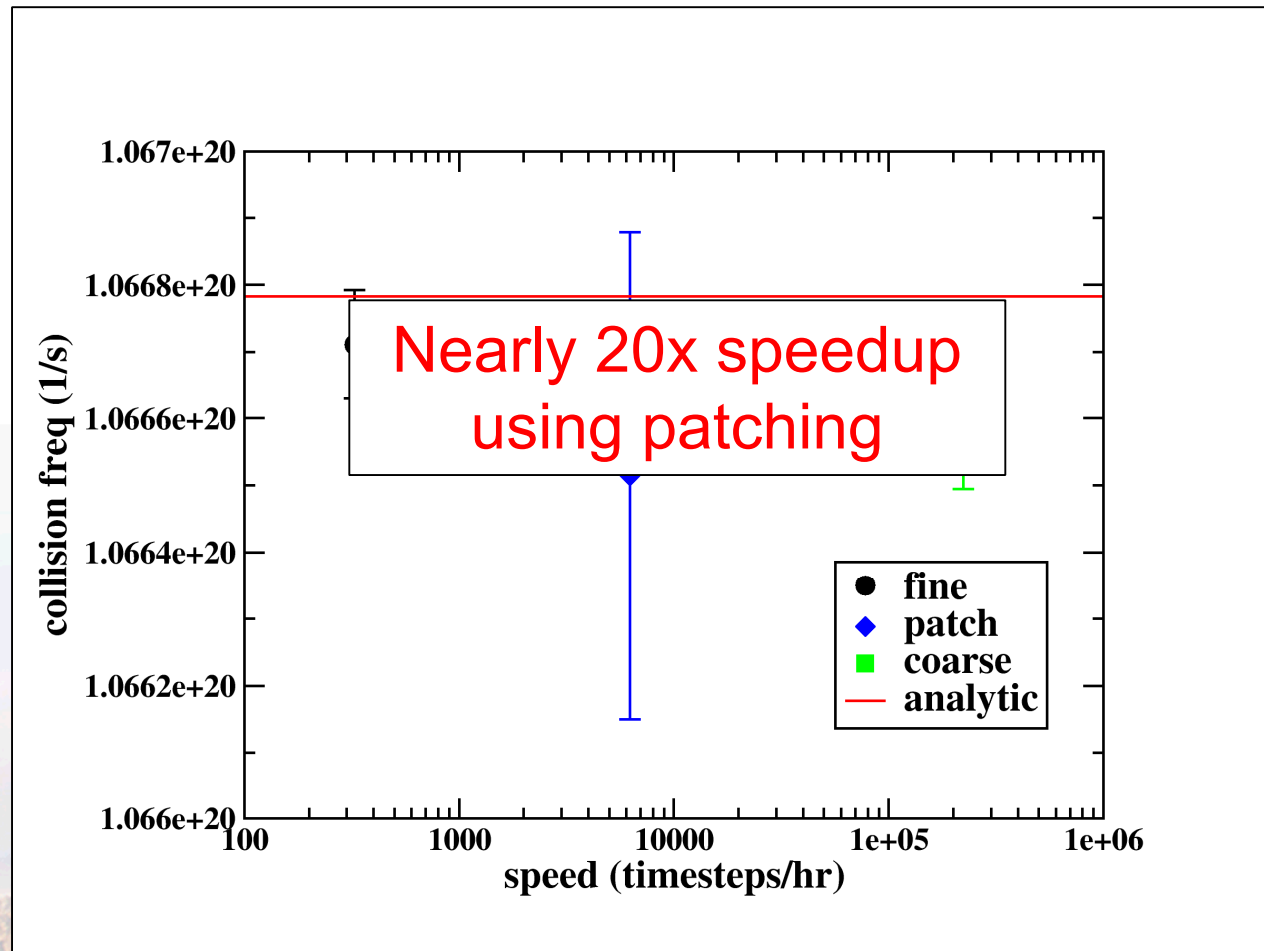
## Average Collision Frequency:

- $f_{\text{analytic}} = 1.0667 \cdot 10^{20} \text{ s}^{-1}$
- $f_{\text{fine}} = 1.0668 \cdot 10^{20} \text{ s}^{-1} \rightarrow 0.007\% \text{ relative error}$
- $f_{\text{patch}} = 1.0665 \cdot 10^{20} \text{ s}^{-1} \rightarrow 0.025\% \text{ relative error}$
- $f_{\text{coarse}} = 1.0666 \cdot 10^{20} \text{ s}^{-1} \rightarrow 0.021\% \text{ relative error}$

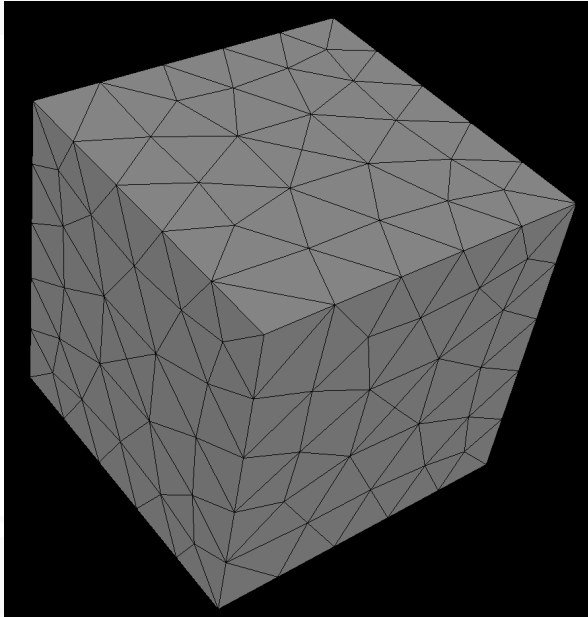


# Results for the HI Test Problem

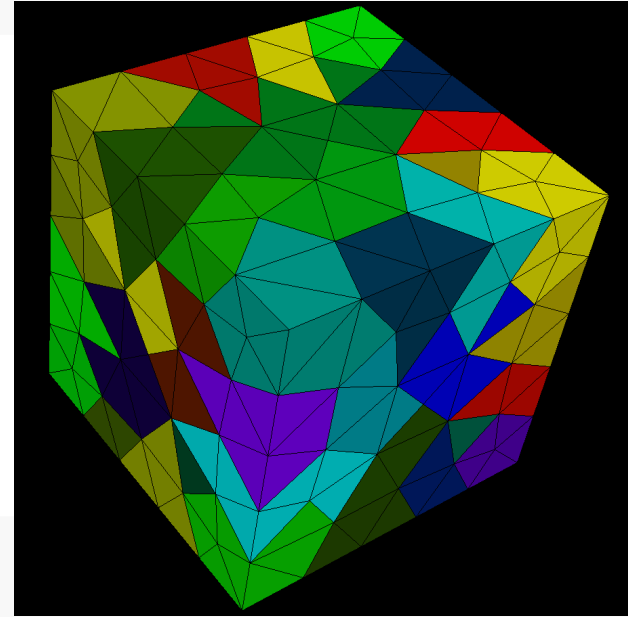
- Error bars represent 95% confidence intervals



## Another 2 Meshes



*fine* mesh  
1,184 elements  
1,223 particles



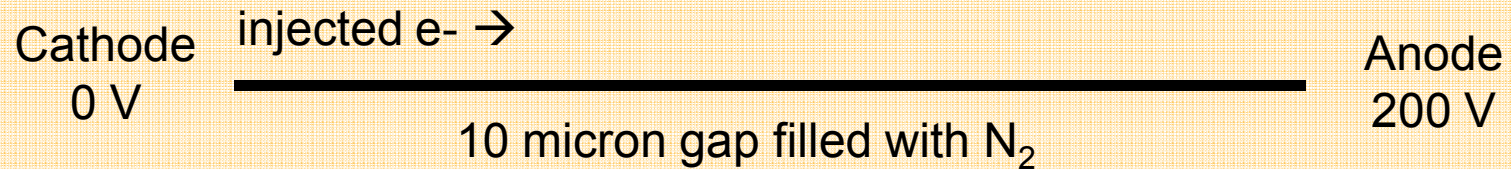
*patched* mesh  
64 patches  
1,223 particles

- Same HI test problem (ran for 2000 timesteps)
- Patched mesh runs nearly **4x** faster (with virtually the same accuracy)
- Global average (0D problem) may hide errors





# 1D Inelastic Collision Problem

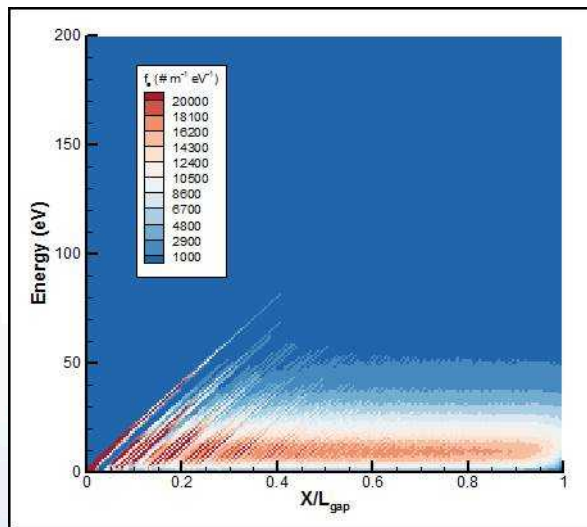


- Electrons gain energy due to the field
- Electrons lose energy due to inelastic collisions with  $N_2$  particles (fixed in space)
- Vary  $N_2$  particle weight so that there are approximately:
  1. 10 computational particles per cell
  2. 0.1 computational particles per cell
- About  $10 \lambda_{mfp}$  across the gap

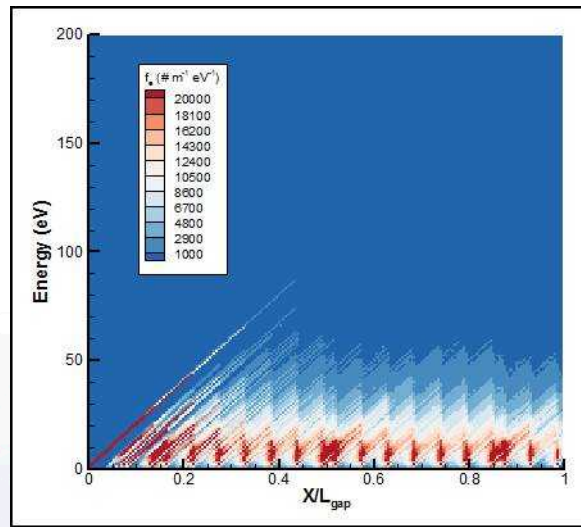


# Electron Energy Distribution Function (EEDF)

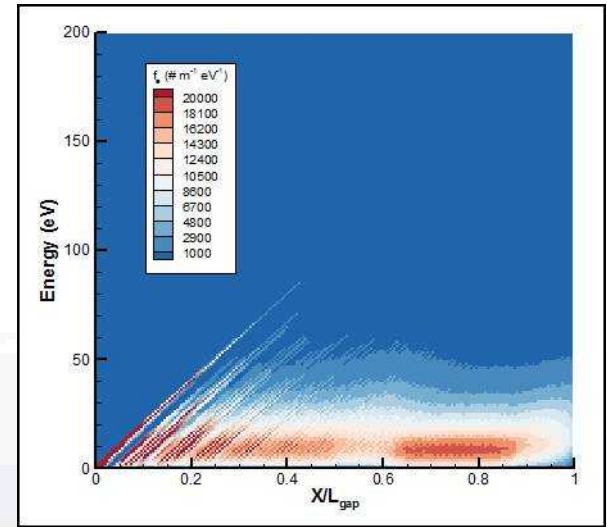
10 N<sub>2</sub> particles  
per cell



0.1 N<sub>2</sub> particles  
per cell



0.1 N<sub>2</sub> particles per cell  
with patching

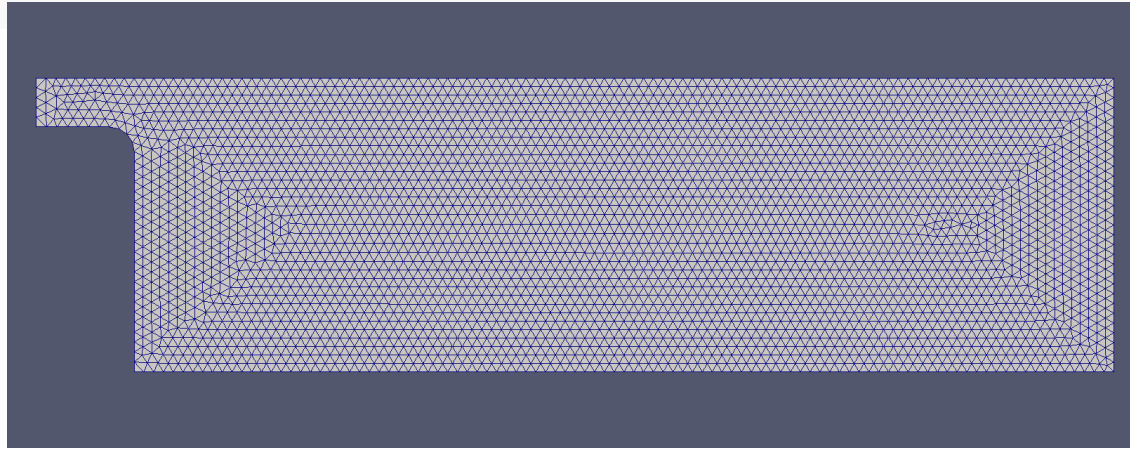


- Using patching gives a more realistic EEDF (crucial for simulating accurate breakdown voltages)
- Patching allows one to use fewer N<sub>2</sub> particles



## 2D Vacuum Gap Breakdown

Anode  
350 V

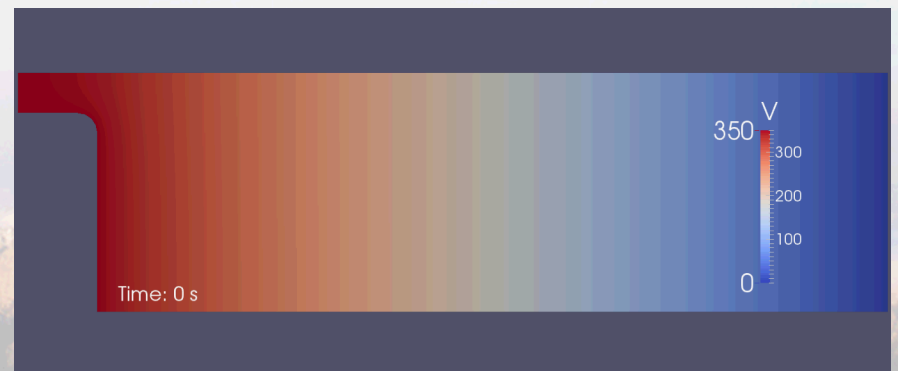


Cathode  
0 V

- Air injected at high velocity and high temperature from the anode
- Low density electrons injected from the cathode
- Air ionizes and eventually will form plasma and break the gap



# Species Densities ( $\text{m}^{-3}$ )





# Dynamic Sizing of DSMC Collision Cells

Time: 0 s

Time: 0 s

4e+025 N2  
3e+25  
2e+25  
1e+25  
0

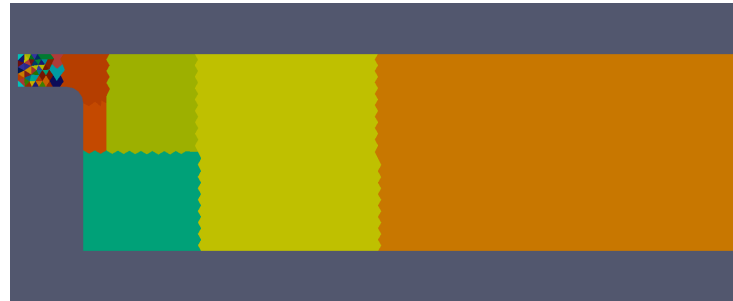
Time: 0 s



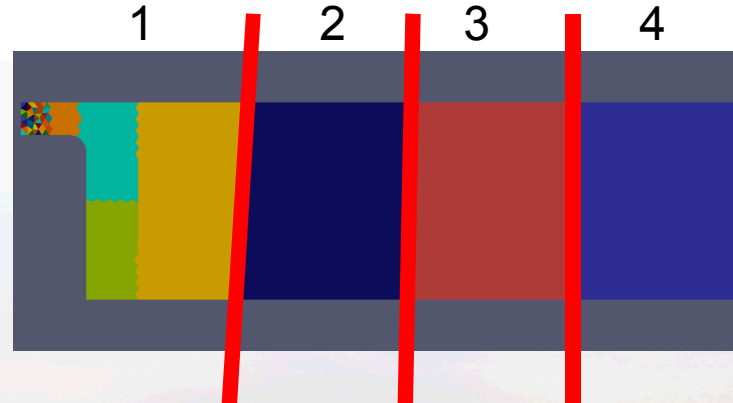
# Parallelization

- Currently each processor uses an independent oct-tree (simple):

1 processor:



4 processors:



- Issue: can lead to narrow/small cells (can happen with 1 proc too)
- In the future, could use a more robust method like inter-processor communication with ghosting, load balancing, etc.





# Conclusions

---

- **Developed a new method (patching) to enhance hybrid PIC-DSMC simulations**
- **Using fewer particles with patching gives similar accuracy and uncertainty as using a fine mesh with many particles**
- **Allows one to dramatically speed up the simulation if the PIC mesh is too small for DSMC collisions**
- **Allows one to dynamically adjust the size of DSMC collision cells on the fly (based on mean free path)**





# Sandia Projects

- Been at Sandia since August, 2012

## Projects:

- **LAMMPS (molecular dynamics) code development:** Added and improved long-range electrostatic methods
- **Modeling of a dipolar fluid in an external electric field** (uses LAMMPS)
- **Advanced force field development (SNAP)** (uses LAMMPS, working to develop a quantum accurate  $\text{SiO}_2$  force field)
- **Aleph (PIC-DSMC) code development:** Added an automatic mesh coarsening method for particle interactions (based on oct-tree algorithm)
- **Modeling of triggered vacuum gap switches**

Thanks to Paul Crozier for being a great mentor during my time at Sandia





# Why do we care about LRE?

LRE = long-range electrostatics

- Electrostatics are extremely important in many atom-level (and coarser) models.
- Long-range part usually cannot be neglected in molecular simulations (simple cutoff can lead to artifacts).
- Large fraction of compute cycles are used in LRE calculations.
- LRE calculations represent a scaling bottleneck in many MD calculations.



Parts of the LAMMPS long-range electrostatic work were done in collaboration with Paul Crozier, Steve Plimpton, and Stephen Bond

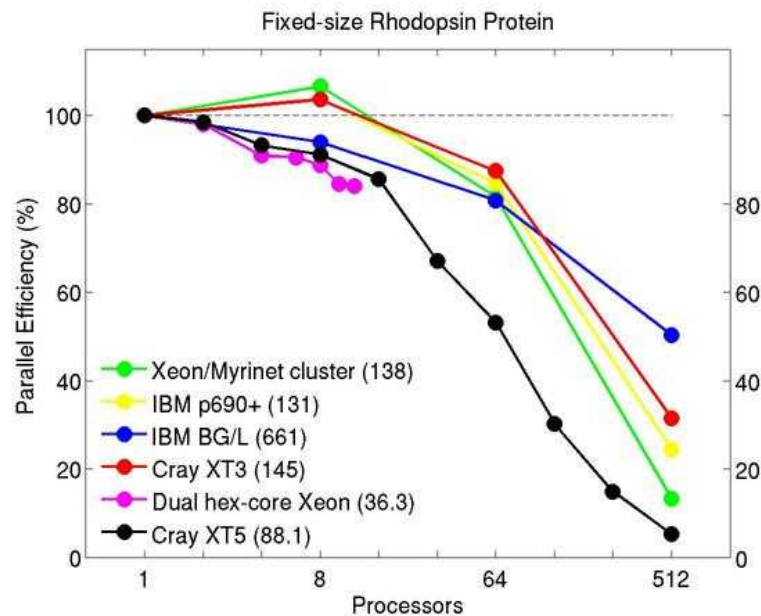


Sandia National Laboratories

# The motivation: FFTs don't scale very well

(and HPC core counts are growing quickly)

- LAMMPS originally had two methods for computing long-range electrostatics: Ewald and particle-particle/particle-mesh (PPPM)
- Ewald summation is fastest for small systems (or very high accuracy), but expensive for large systems
- PPPM relies on FFTs, which don't scale well on many processors:



<http://lammps.sandia.gov/bench/rhodo.fixed.jpg>



Sandia National Laboratories

# Multilevel-summation method (MSM) background

- Multi-grid method (but not iterative); split potential and approximate the slowly varying part on a hierarchy of grids
- No FFTs are required, so the communication cost of MSM is expected to scale better than PPPM on large core counts
- PPPM scales with number of atoms as  $O(N \log N)$  while MSM scales as  $O(N)$
- MSM may be faster for large problems running on large core counts

D. J. Hardy, Ph.D. thesis, University of Illinois at Urbana-Champaign (2006).

D. J. Hardy, J. E. Stone, and K. Schulten, *Parallel Comput.* **35**, 164 (2009).





# MSM Algorithm pieces

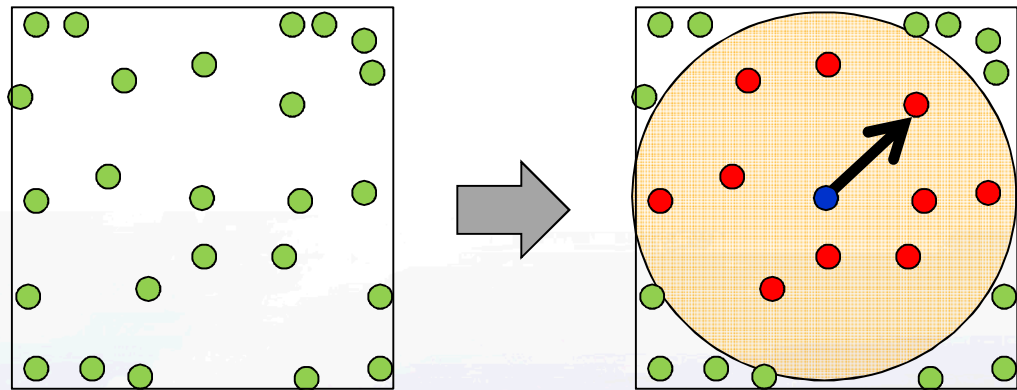
1. Short-range part
2. Anterpolation
3. Direct sum
4. Restriction
5. Prolongation
6. Interpolation





# Algorithm pieces

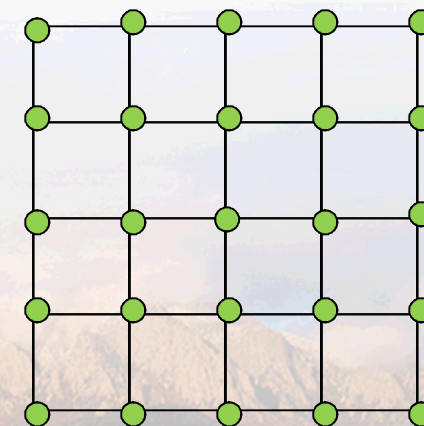
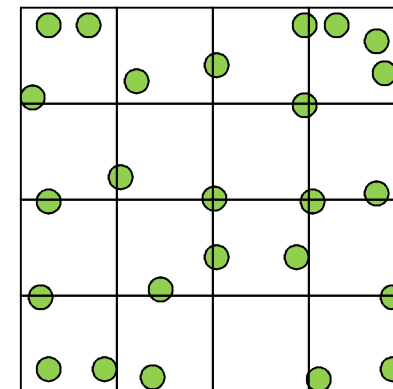
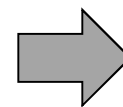
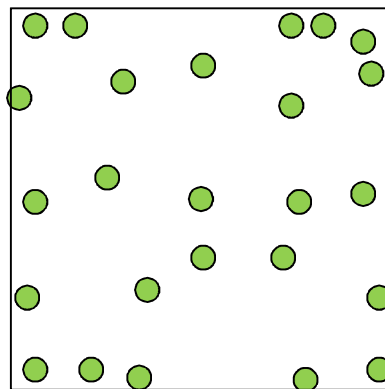
1. **Short-range part**
2. Anterpolation
3. Direct sum
4. Restriction
5. Prolongation
6. Interpolation



**Particles interact within a spherical cutoff to get short-range forces, energy, and pressure**

# Algorithm pieces

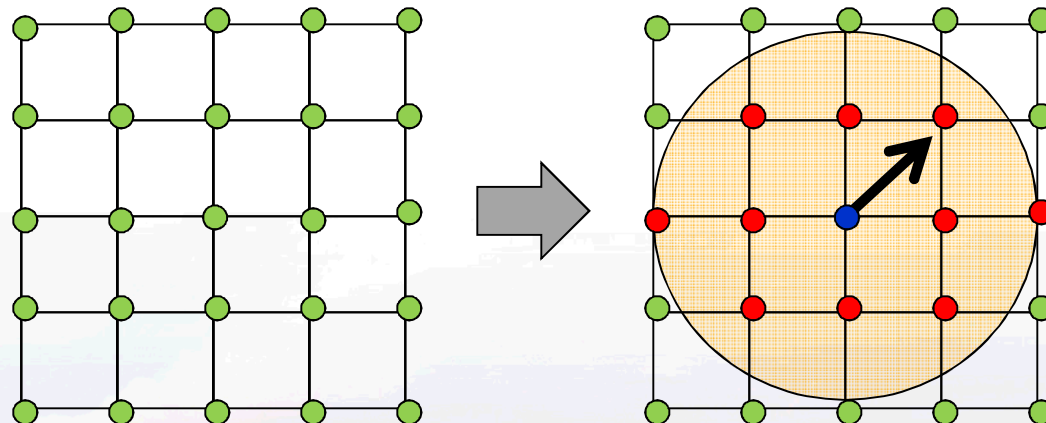
1. Short-range part
2. **Anterpolation**
3. Direct sum
4. Restriction
5. Prolongation
6. Interpolation



Interpolate charges from atoms to the finest mesh

# Algorithm pieces

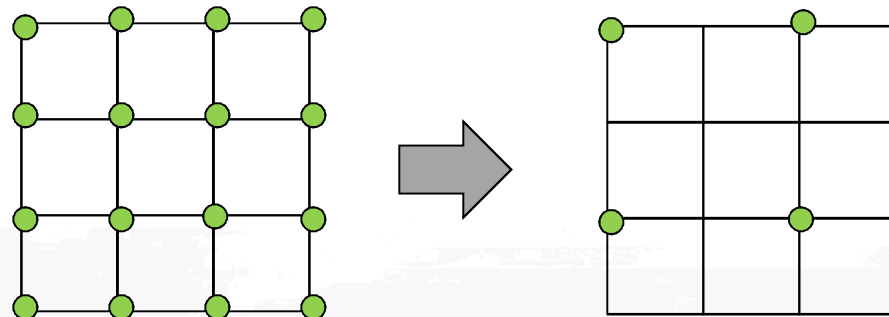
1. Short-range part
2. Anterpolation
3. **Direct sum**
4. Restriction
5. Prolongation
6. Interpolation



**Mesh points interact within a spherical cutoff to get long-range electric field, energy, and pressure**

# Algorithm pieces

1. Short-range part
2. Anterpolation
3. Direct sum
4. **Restriction**
5. Prolongation
6. Interpolation



Interpolate charge from finer mesh to coarser mesh

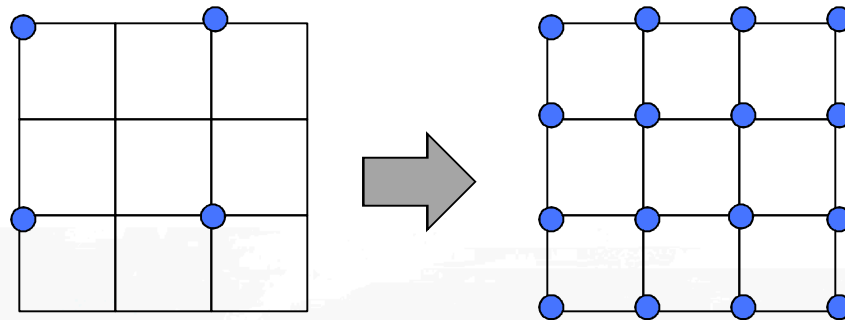
Repeat steps 3 and 4 until finished on coarsest mesh





# Algorithm pieces

1. Short-range part
2. Anterpolation
3. Direct sum
4. Restriction
5. **Prolongation**
6. Interpolation

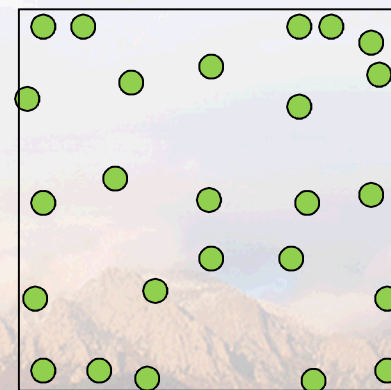
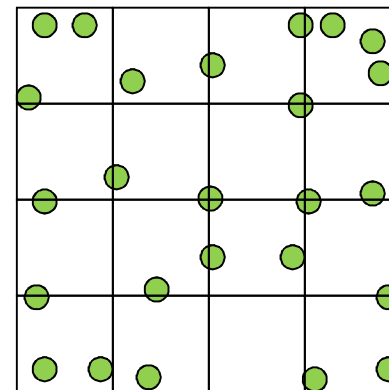
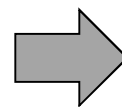
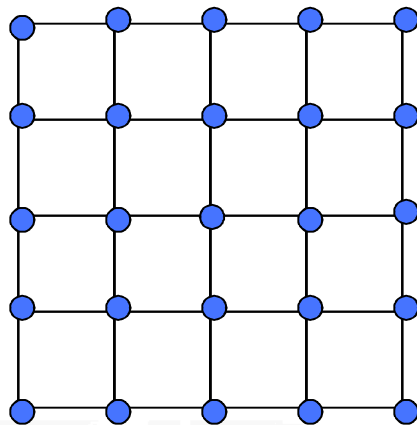


Interpolate electric field from coarser mesh to finer mesh

Repeat step 5 until the finest mesh is reached

# Algorithm pieces

1. Short-range part
2. Anterpolation
3. Direct sum
4. Restriction
5. Prolongation
6. **Interpolation**

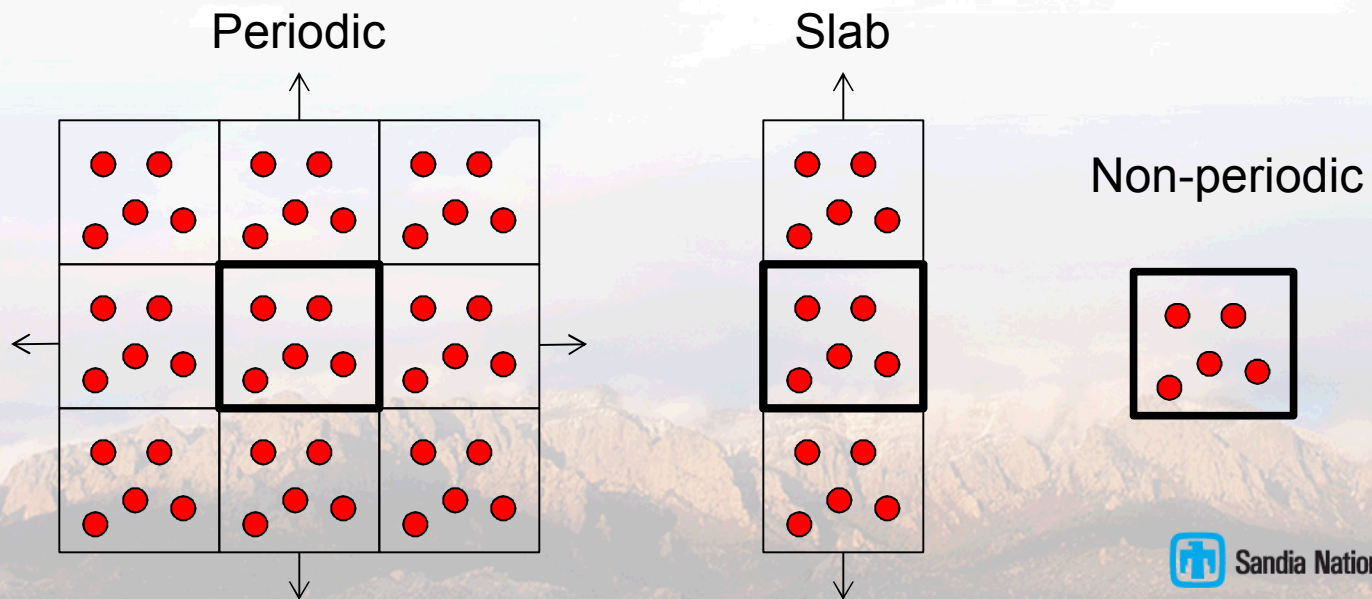


Compute force from electric field on finest mesh and  
back-interpolate force from mesh to atoms



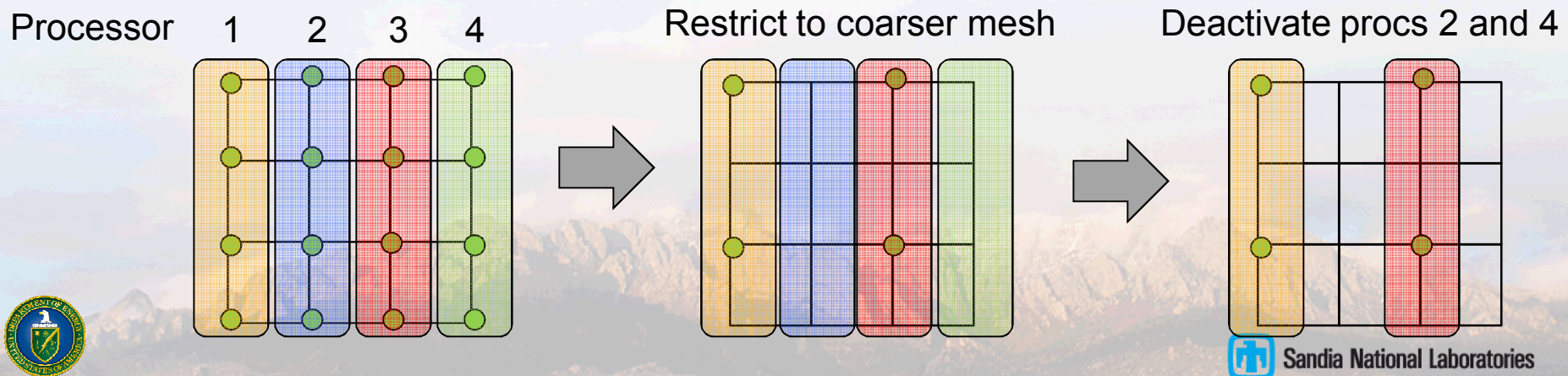
# Non-Periodic BCs

- MSM works for both periodic and non-periodic boundary conditions
- Ewald and PPPM only work for periodic or slab (periodic in  $x$  and  $y$  and non-periodic in  $z$ ) boundary conditions



# Parallelization Strategy

- **Challenge:** lots of work on finest grid, very little work on coarsest grid
- **Use same domain-decomposition layout on all levels (simple)**
- **Inactive processors don't participate in MPI communication routines**
- **Use neighbor point-to-point communication for fine grids**
- **Use MPI AllReduce for coarse grids**







# Enhancements to MSM

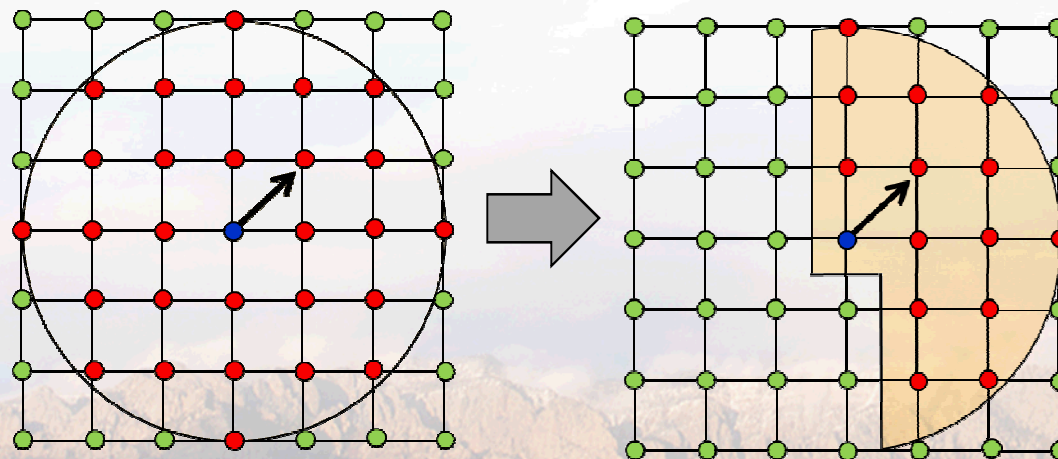
---

- **Error estimator (important for comparing to other methods like PPPM)**
- **Pressure calculation**
- **Added heuristic to estimate optimal parameters, including automatic adjustment of Coulombic cutoff (based on work by Hardy)**
- **Per-atom energy/virial**
- **Fast scalar pressure**
- **OMP threaded version of MSM (Axel Kohlmeyer)**



# Improving Single-Core Performance

- Use hemisphere (instead of full sphere) for direct sum interactions to avoid double computations
- Using a hemisphere can also (sometimes) reduce the amount of communication needed
- Added various other code optimizations
- Compared to Hardy's NAMD-lite code, LAMMPS MSM was 60% faster for periodic and 25% faster for non-periodic (1 processor, two point-charges in a box, order 4)



[David J. Hardy, *NAMD-Lite*, <http://www.ks.uiuc.edu/Development/MDTools/namdlite/>, University of Illinois at Urbana-Champaign, 2007.]



Sandia National Laboratories



# Fast (Scalar) Pressure Calculation

- Calculation of the 6-component pressure tensor is expensive with MSM (increases cost by ~2x)
- Often only scalar pressure [i.e.  $\frac{1}{3}(P_{xx} + P_{yy} + P_{zz})$ ] is needed
- For Coulombic systems, can use a virial “trick” to relate energy to scalar pressure (much cheaper)
- For SPC/E system, reduces overall cost by 20% (short-range part has some overhead)
- Can use scalar pressure to run isotropic barostat



# LRE speed and scalability tests

## Chama

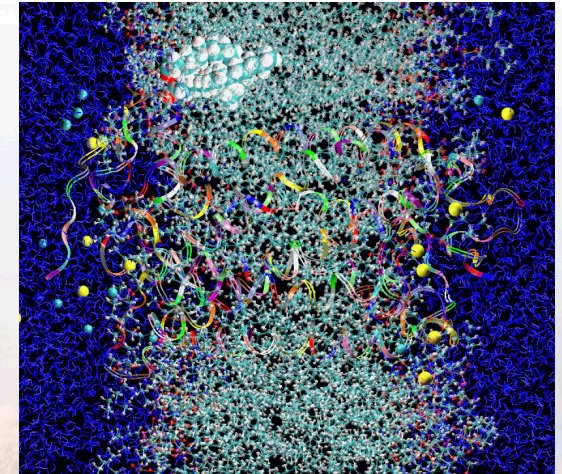
- Sandy-Bridge (2.6 GHz)
- 2012
- 16 cores/node
- 1,232 nodes
- 19,712 cores
- Infiniband 4X QDR, Fat Tree, Qlogic

## Redsky

- Nehalem (2.93 GHz)
- 2009
- 8 cores/node
- 2,816 nodes
- 22,528 cores
- Infiniband 4X QDR, 3D Torus, Mellanox

## Rhodopsin benchmark

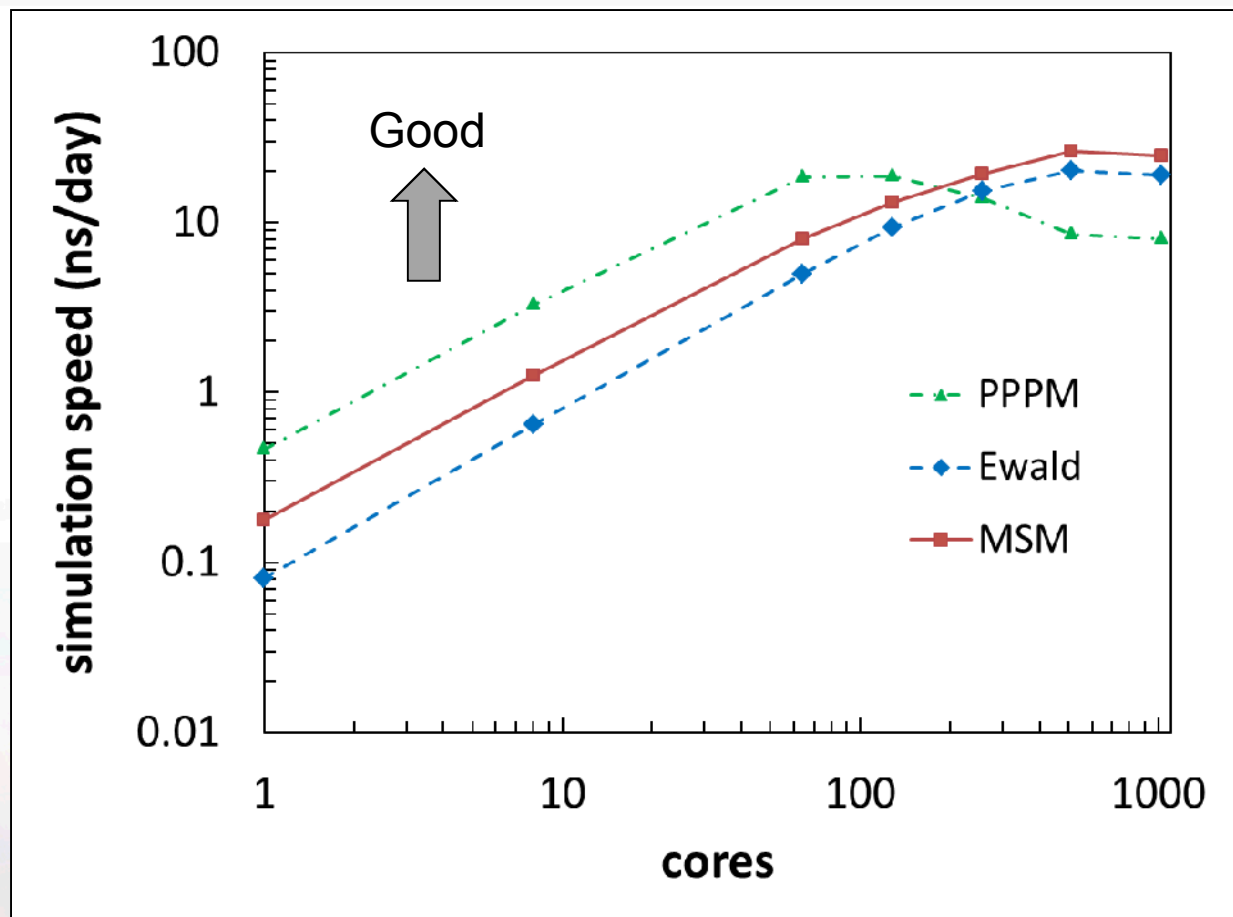
- NVT dynamics
- 1e-4 relative accuracy
- 32k atoms, replicable
- 2 fs timestep size





## Redsky, 32k atoms, $10^{-4}$ accuracy

- PPPM is fastest at low core count
- MSM is fastest at high core count
- MSM scales better than PPPM since it doesn't rely on FFTs
- Maxes out at 26 ns/day

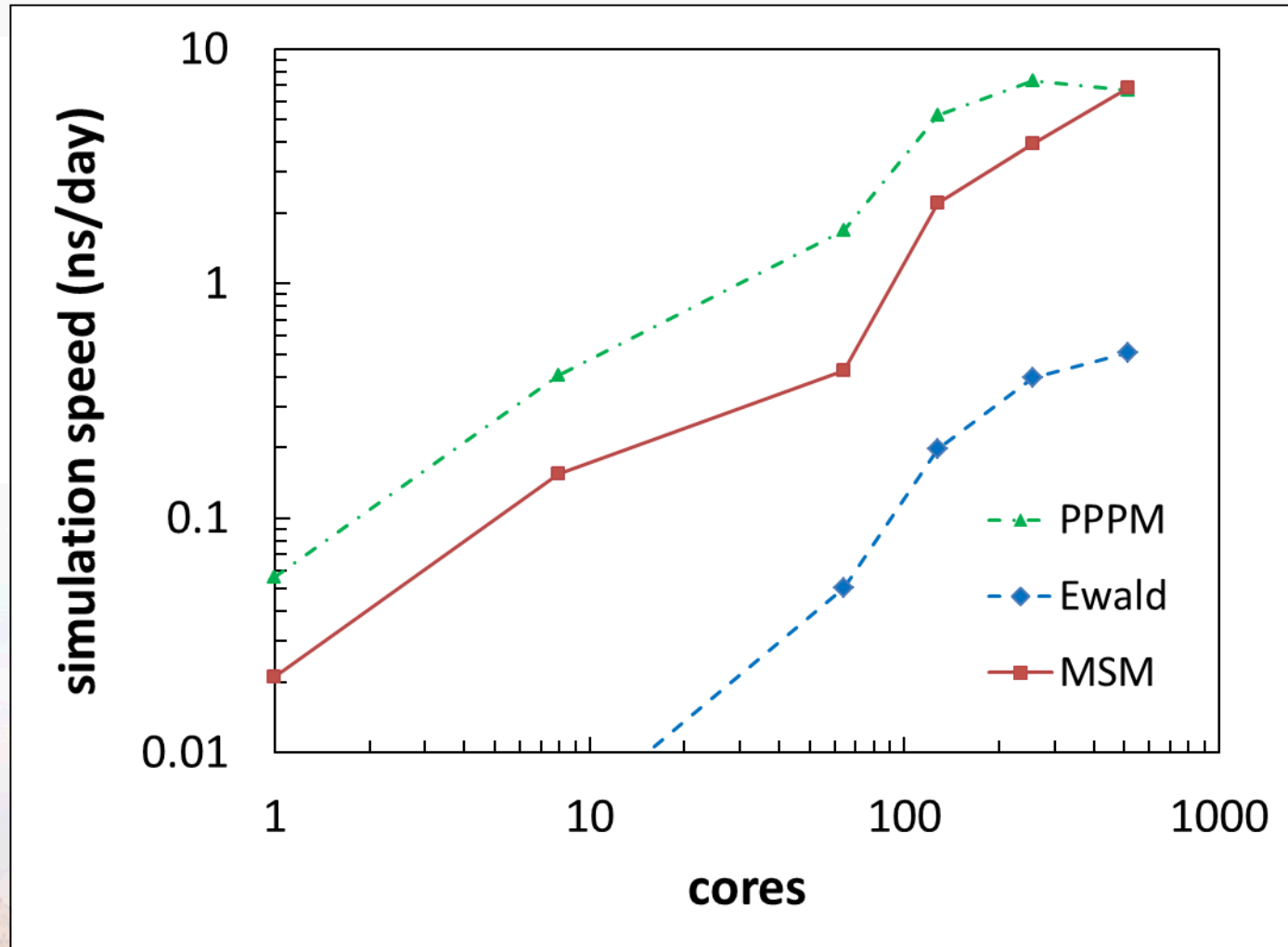


10 Angstrom cutoff, default parameters, MSM order 10



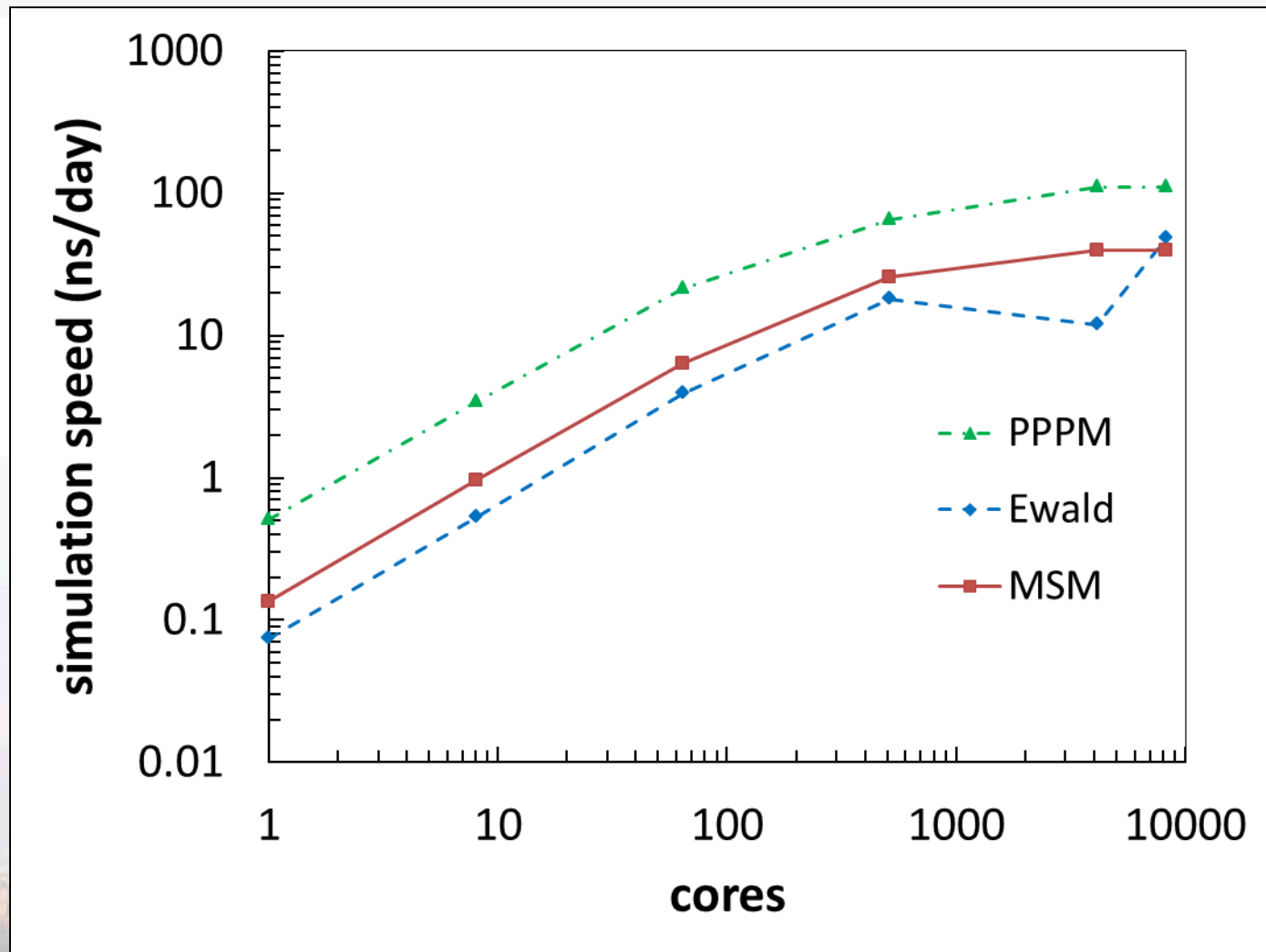
## Redsky, 256k atoms, $10^{-4}$ accuracy

- 8x as many atoms as before
- Ewald chokes
- PPPM FFT bottleneck pushed out to larger core count
- Scaling: MSM  $O(N)$ , PPPM  $O(N \log(N))$ , Ewald  $O(N^{1.5})$



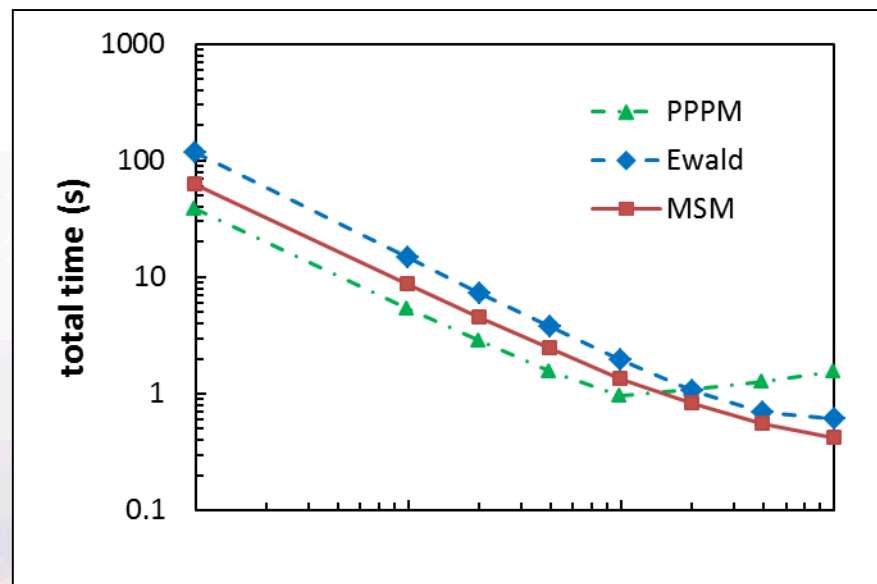
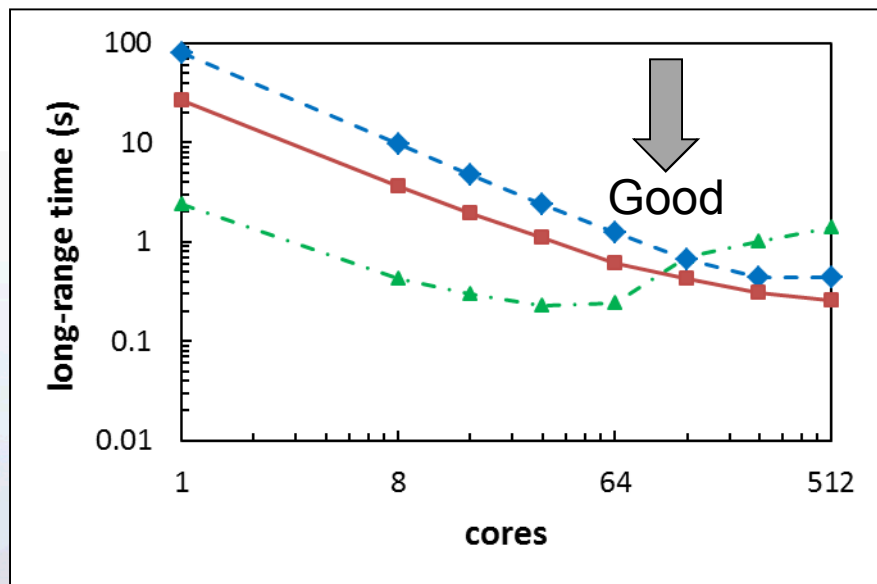
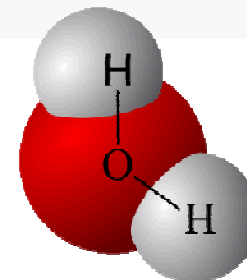
## Chama, 32k atoms, $10^{-4}$ accuracy

- Chama gives better overall performance
- PPPM wins, but we can't expect more scaling
- Maxes at 110 ns/day



# SPC/E Water Benchmark

- 36,000 atoms (strong scaling on Redsky)
- NVT, pressure computed every 50 timesteps
- 1e-3 accuracy

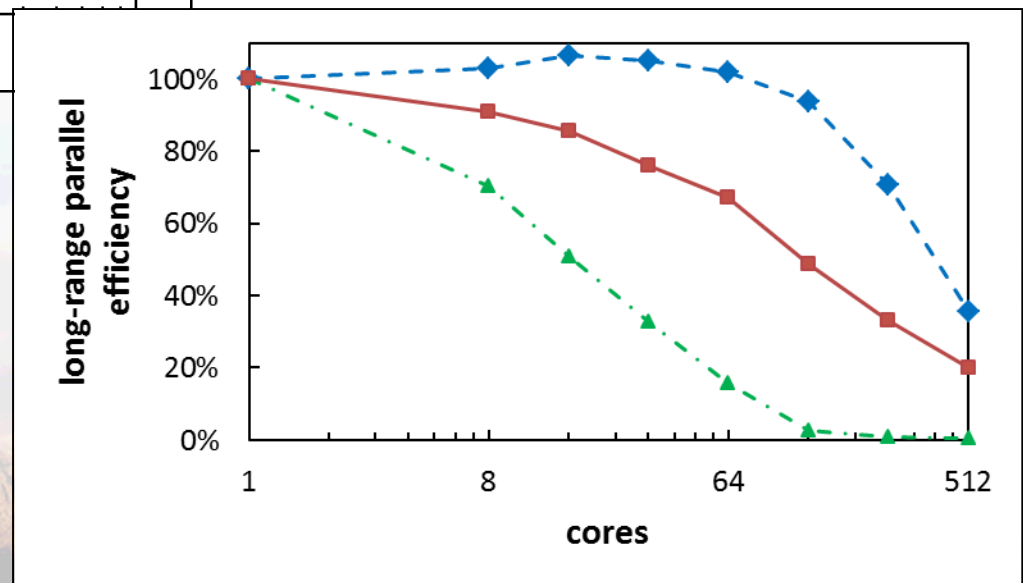
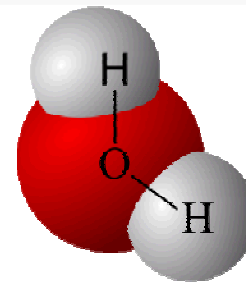
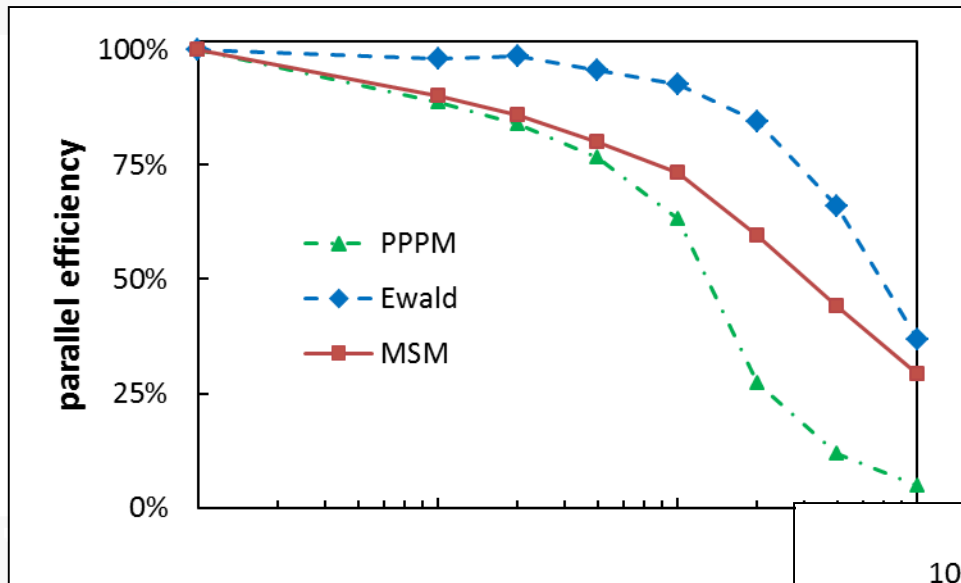


10 Angstrom cutoff, default parameters, MSM order 8





# SPC/E Water Benchmark





# Other Enhancements to LRE in LAMMPS

- Per-atom energy/virial for kspace
- Compute group/group for PPPM and Ewald
- Triclinic for kspace
- Ewald/disp for point-dipoles
- Staggered PPPM (up to 4x faster for high accuracy)

Good for PPPM on large core counts:

- Fix verlet/split
- 2 FFT PPPM





# Conclusions

## ■ MSM works well for:

- large core counts where many-to-many communication is expensive
- non-periodic BCs
- lower accuracy ( $1e-4$  and below)
- large numbers of atoms (at least in theory due to better  $O(N)$  scaling)

## ■ Ewald works well for small systems

## ■ PPPM is good for many practical systems and is hard to beat





**Thank You**

**Questions or Comments?**



**Sandia National Laboratories**