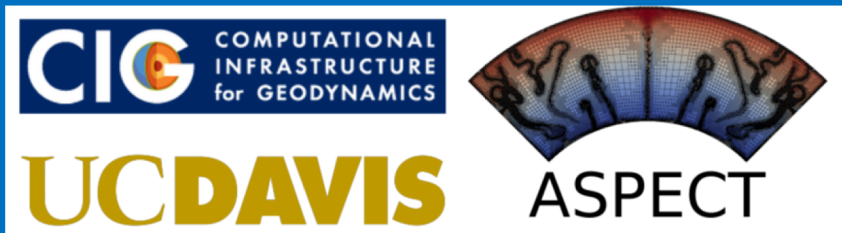
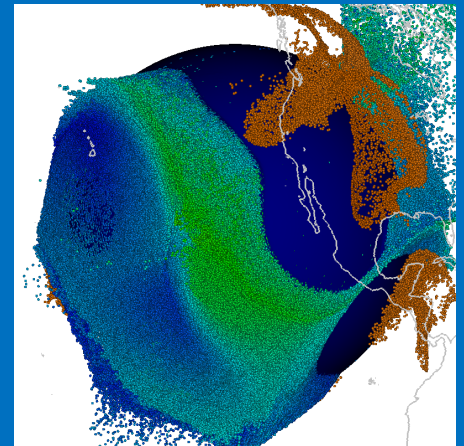
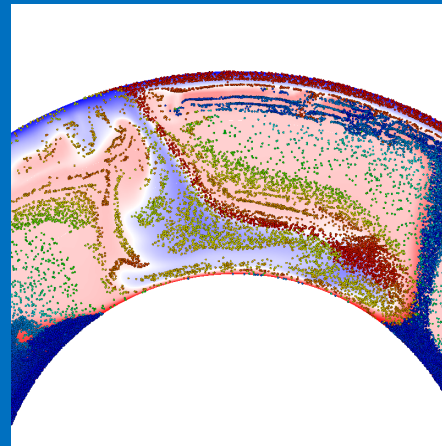
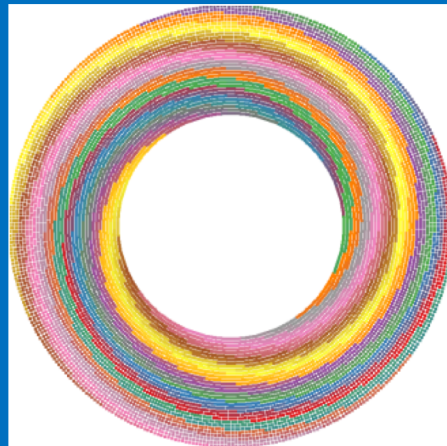
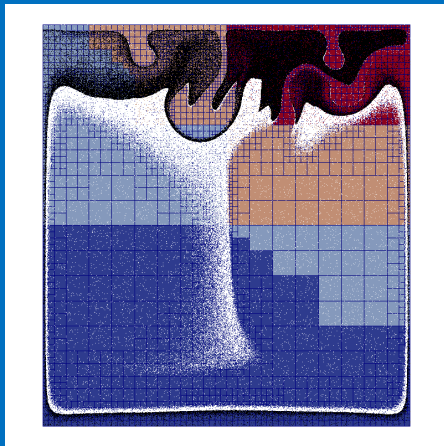


# ACCURATELY UTILIZING PARTICLE-IN-CELL METHODS FOR ADAPTIVELY REFINED FINITE-ELEMENT MODELS



**Rene Gassmoeller**

University of California, Davis

With work by: Eric Heien, Elbridge Gerry Puckett,  
Harsha Lokavarapu, Wolfgang Bangerth

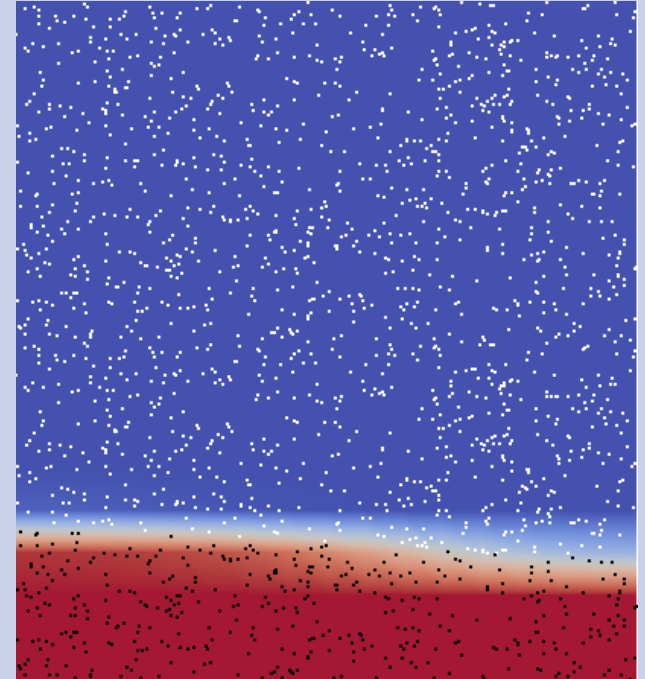
# CONCEPTS OF PIC METHODS

$$-\nabla \cdot [2\eta \dot{\boldsymbol{\epsilon}}(\mathbf{u})] + \nabla p = \rho \mathbf{g} \quad (1)$$

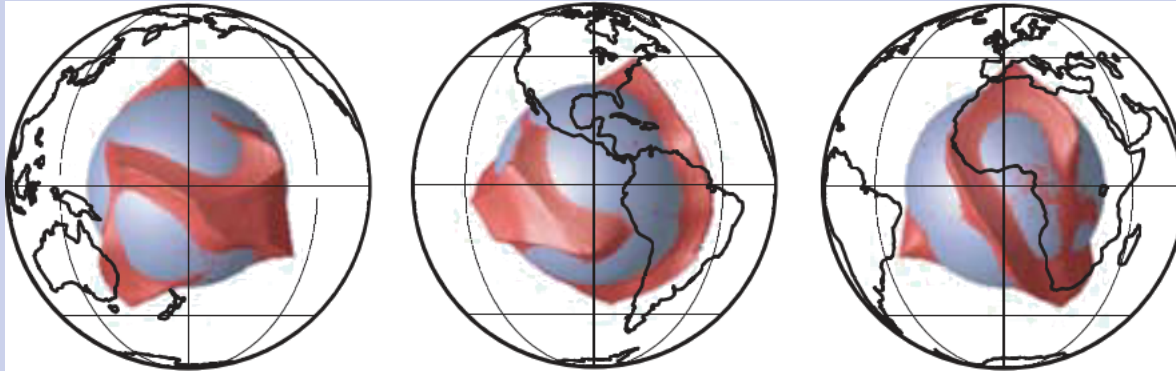
$$\nabla \cdot \mathbf{u} = 0 \quad (2)$$

$$\frac{\partial q}{\partial t} + \mathbf{u} \cdot \nabla q - \nabla \cdot (D \nabla q) = 0 \quad (3)$$

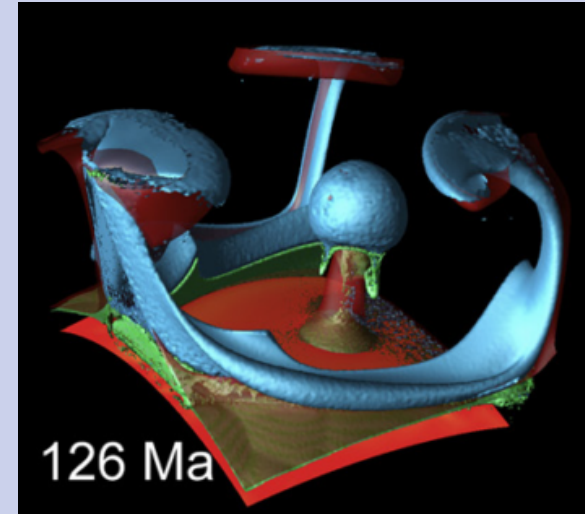
- Method to advect property on discrete particles (/tracers/markers)
- Transforms the PDE in (3) to a set of (non-coupled) ODE's
- Many variants and evolutions (e.g. Particle-in-cell, Evans & Harlow 1957; Marker-and-cell, Harlow & Welch 1965; Marker-in-cell, Gerya 2003; ...)



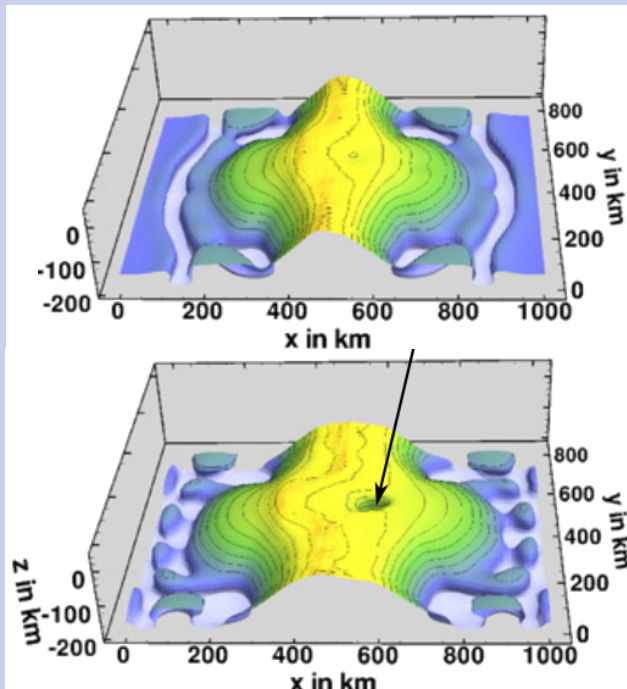
# EXAMPLE APPLICATIONS IN GEODYNAMICS



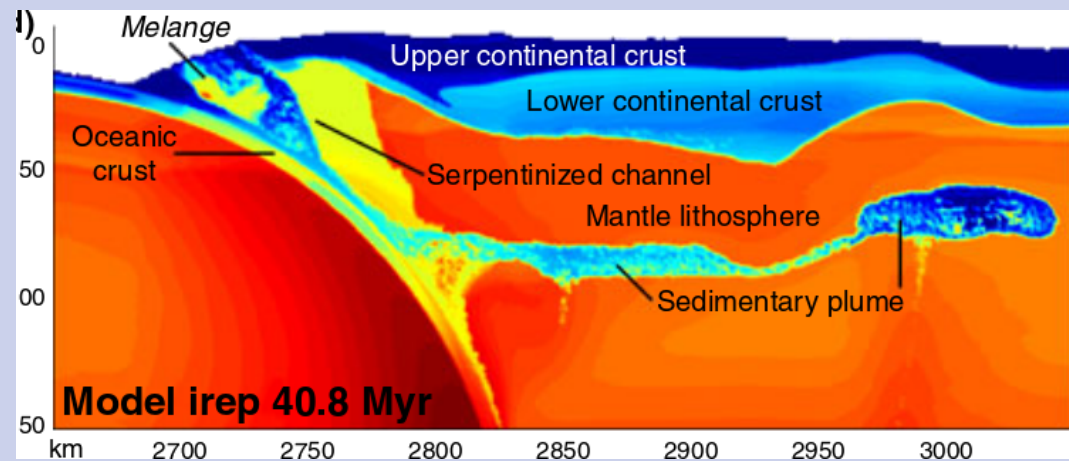
McNamara  
& Zhong, 2005



126 Ma  
Tackley, 2011



Brune et al., 2013



Gerya & Meilick, 2011

# MODERN PARTICLE METHODS:



Challenges when applying Particle methods to modern finite-element codes:

- Quantifying accuracy and convergence behavior
- Load balancing strategies in adaptively refined meshes
- Geometry independent particle-cell search

# ACCURACY: CONVERGENCE BEHAVIOR



Known from previous studies:

- Increasing the number of particles per cell converges towards a solution (known from other fields, geodynamics: Tackley & King, 2003)
- Particles do not decrease convergence rate for second order accurate methods if interpolation is accurate enough (Thielmann et al, 2014)
- Suggestion: Particles limit the accuracy of the velocity solution to at most second order (Thielmann et al, 2014)

Our goal:

- Quantify the influence of the particles on the accuracy of the solution, in particular in dependence of number of particles per cell (PPC)?
- Is it possible to increase the convergence rate above second order?

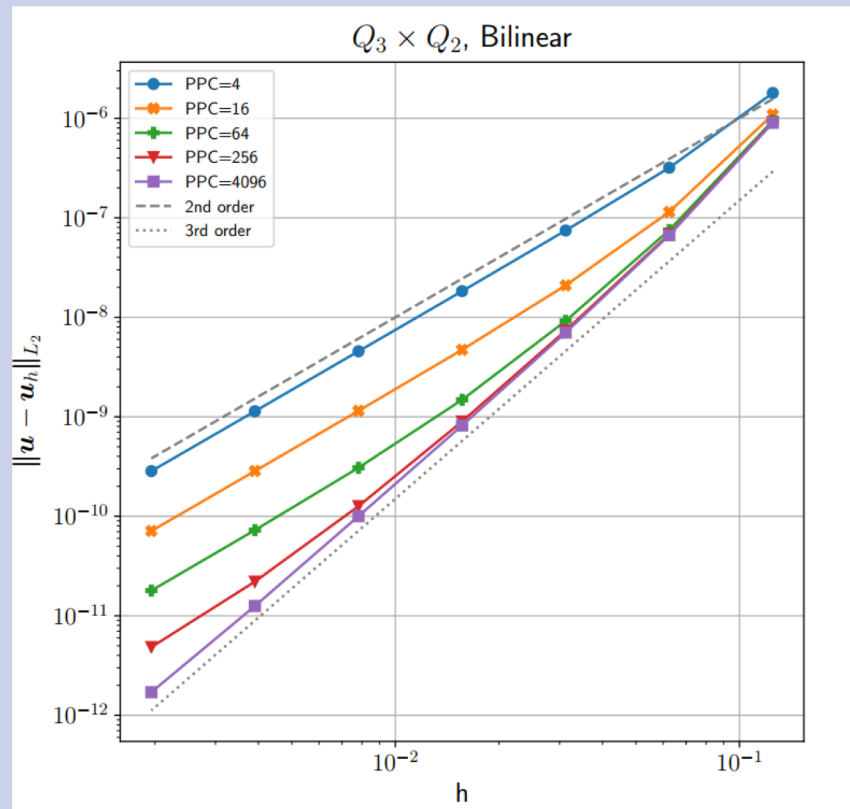
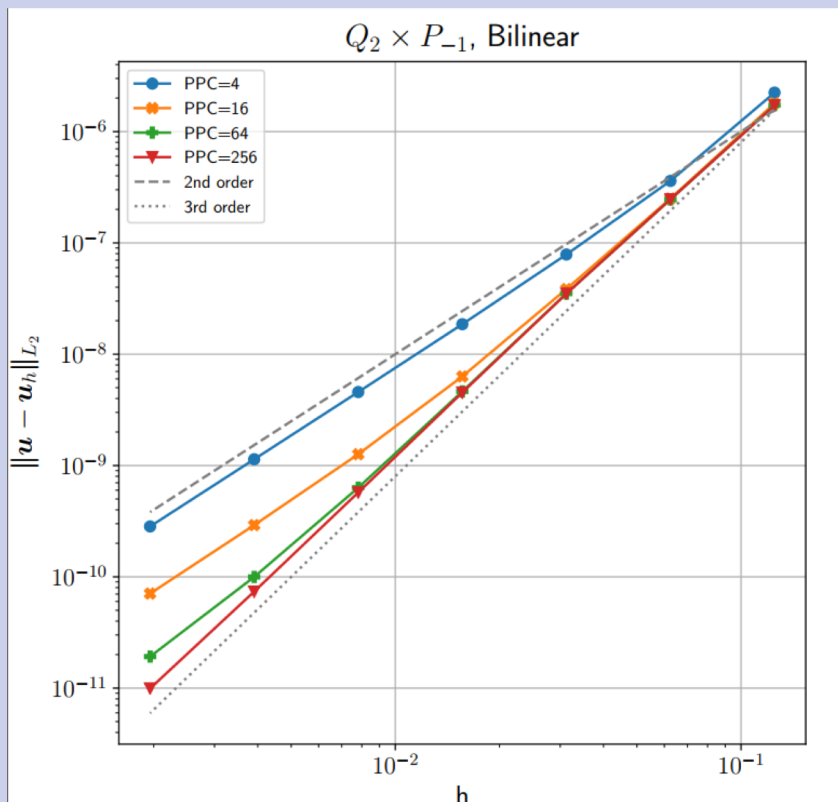
Our approach:

- Use existing instantaneous benchmarks and quantify accuracy for different FEs, different interpolation schemes, and different PPC
- Develop a theoretical understanding of the underlying error sources
- Develop a new time-dependent benchmark with analytical solution

# ACCURACY: CONVERGENCE BEHAVIOR

Using the SolKz (Durez et al., 2011) benchmark:

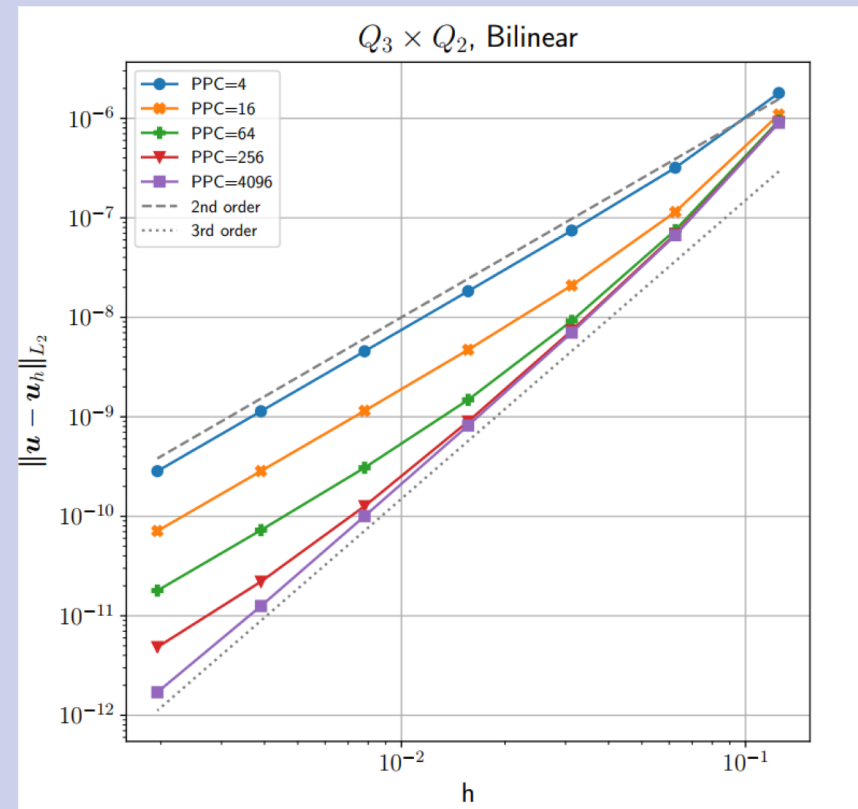
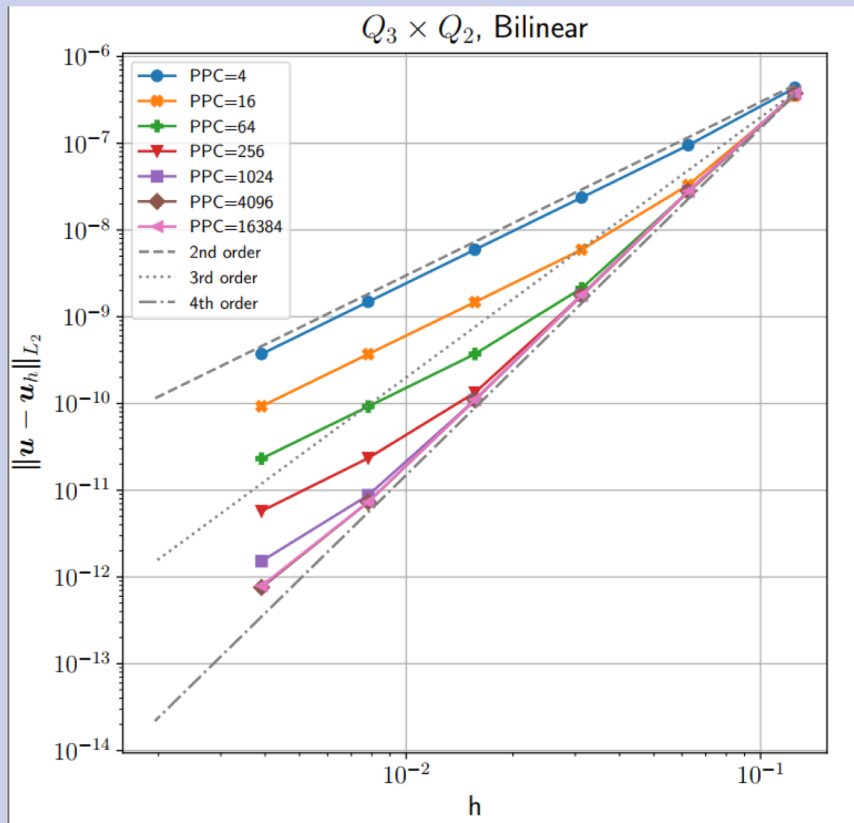
- For constant PPC: 2<sup>nd</sup> order velocity convergence (like Thielmann et al, 2014)
- For increasing PPC with resolution: 3<sup>rd</sup> order velocity convergence
- Even for a Q3xQ2 element: 3<sup>rd</sup> order convergence (expected: 4<sup>th</sup> order)



# ACCURACY: CONVERGENCE BEHAVIOR

Using the SolKz benchmark:

- For an analytic viscosity: up to 4<sup>th</sup> order velocity convergence
- In general: Viscosity on particles limits convergence more strictly



# ACCURACY: THEORETICAL ERROR

Total error = difference between continuous and discretized Stokes operator and properties:

- (1) Error by density approximation
- (2) Error by viscosity approximation
- (3) Error by evaluation at particle locations
- (4) Error by finite element approximation

$$\begin{aligned} & (\eta_0 \|\nabla(\mathbf{u} - \mathbf{u}_h)\|_{L_2}^2 + \|p - p_h\|_{L_2}^2)^{1/2} \\ &= \|\mathcal{L}_\eta(\rho \mathbf{g}) - \mathcal{L}_{\eta_h}^h(\rho_h \mathbf{g})\| \\ &\leq \underbrace{\|\mathcal{L}_\eta(\rho \mathbf{g}) - \mathcal{L}_\eta(P_h \rho \mathbf{g})\|}_{(1)} \\ &\quad + \underbrace{\|\mathcal{L}_\eta(P_h \rho \mathbf{g}) - \mathcal{L}_{P_h \eta}(P_h \rho \mathbf{g})\|}_{(2)} \\ &\quad + \underbrace{\|\mathcal{L}_{P_h \eta}(P_h \rho \mathbf{g}) - \mathcal{L}_{\eta_h}(P_h \rho \mathbf{g})\|}_{(3)} \\ &\quad + \underbrace{\|\mathcal{L}_{\eta_h}(P_h \rho \mathbf{g}) - \mathcal{L}_{\eta_h}^h(\rho_h \mathbf{g})\|}_{(4)}. \end{aligned}$$



# ACCURACY: THEORETICAL ERROR

Velocity error:

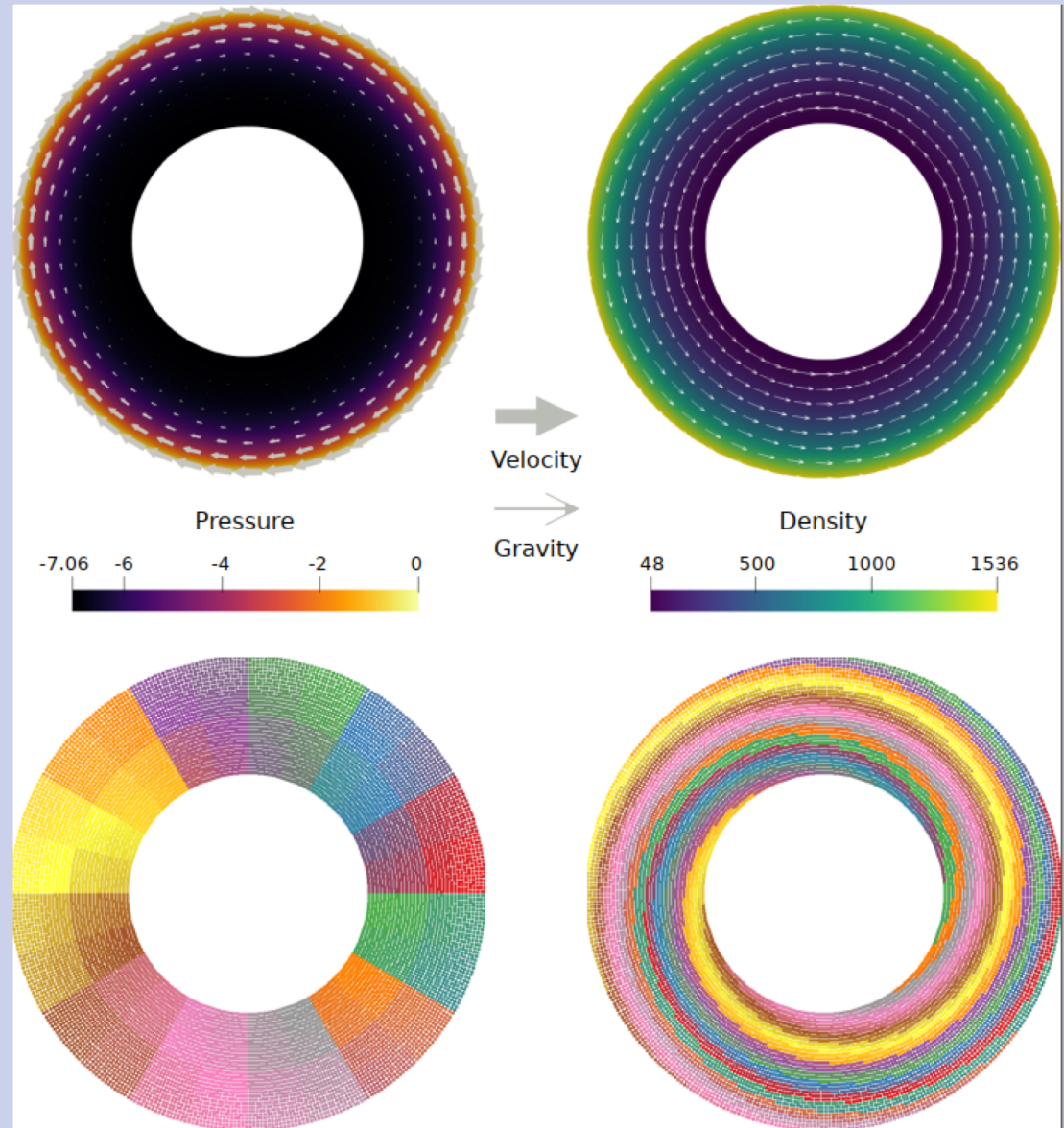
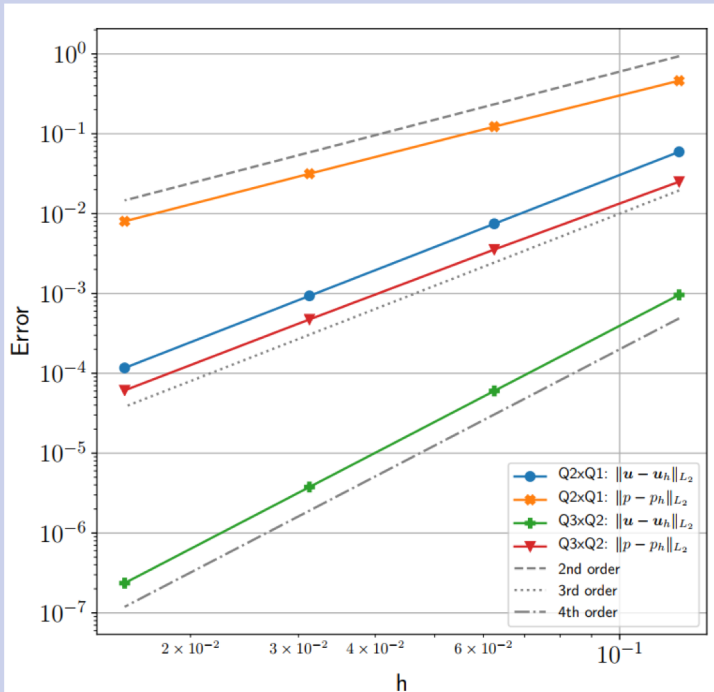
$$\|\mathbf{u} - \mathbf{u}_h\|_{L_2} = \mathcal{O}(h^{r+2}) + \mathcal{O}(h^{r+1}) + \mathcal{O}(h E(h, PPC)) + \mathcal{O}(h^{k+1}).$$

- The convergence order of the interpolation method ( $r$ ) places an upper limit on the velocity accuracy, just like the choice of finite element degree ( $k$ )
- This upper limit depends on whether particles only carry density ( $r+2$ ) or also viscosity ( $r+1$ )
- There is a hard to quantify term  $E$  that depends on PPC and  $h$  (we will experimentally try to estimate this term next)

# ACCURACY: NEW BENCHMARK

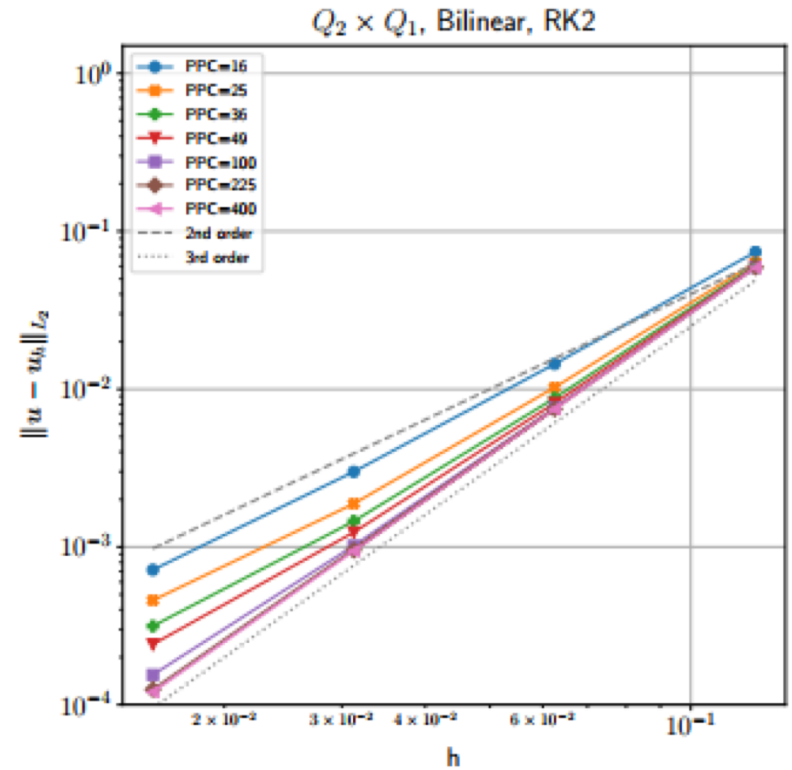
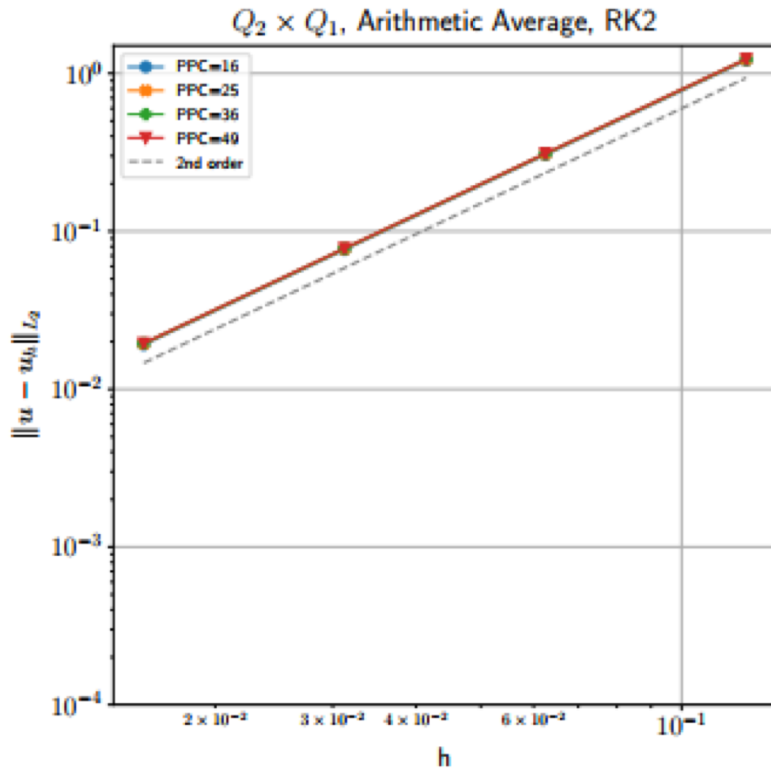
New benchmark: Circular flow

- Analytical solution: Time independent
- Numerical solution : Time dependent error
- Pure FE method reaches design convergence



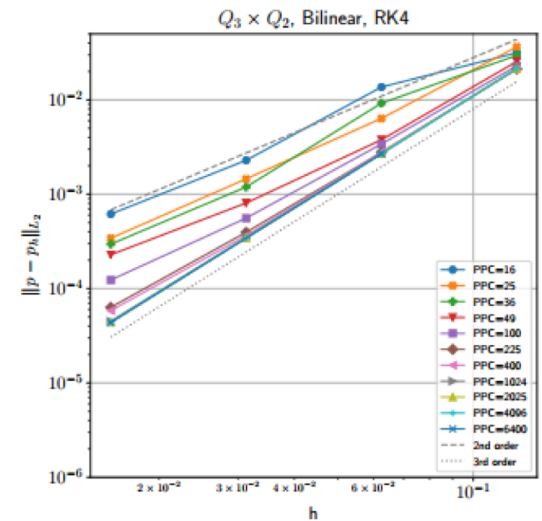
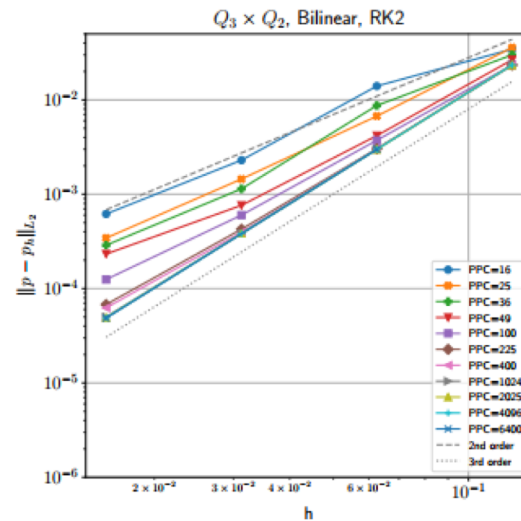
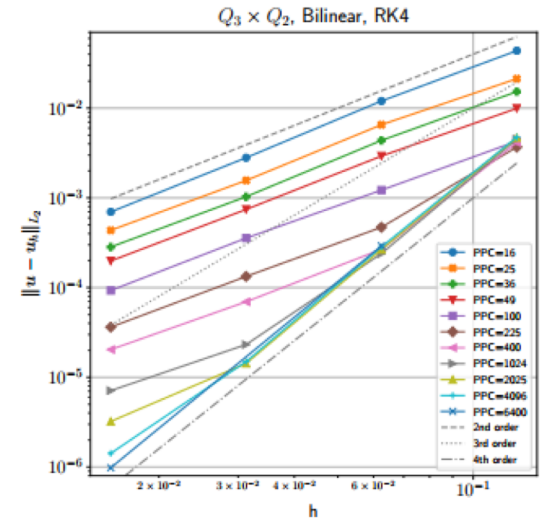
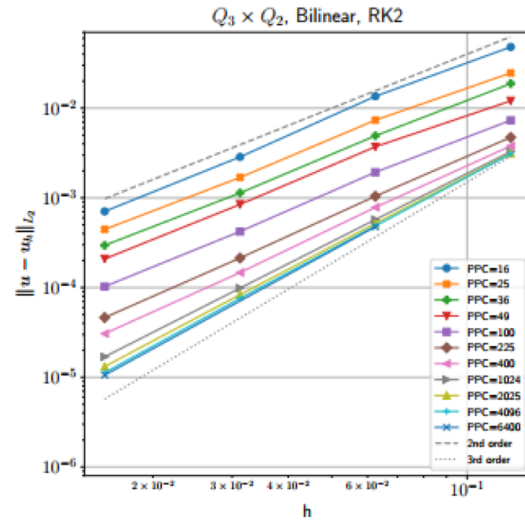
# ACCURACY: CONVERGENCE BEHAVIOR

A comparison of different interpolation methods (arithmetic average vs bilinear least squares approximation) shows expected results:



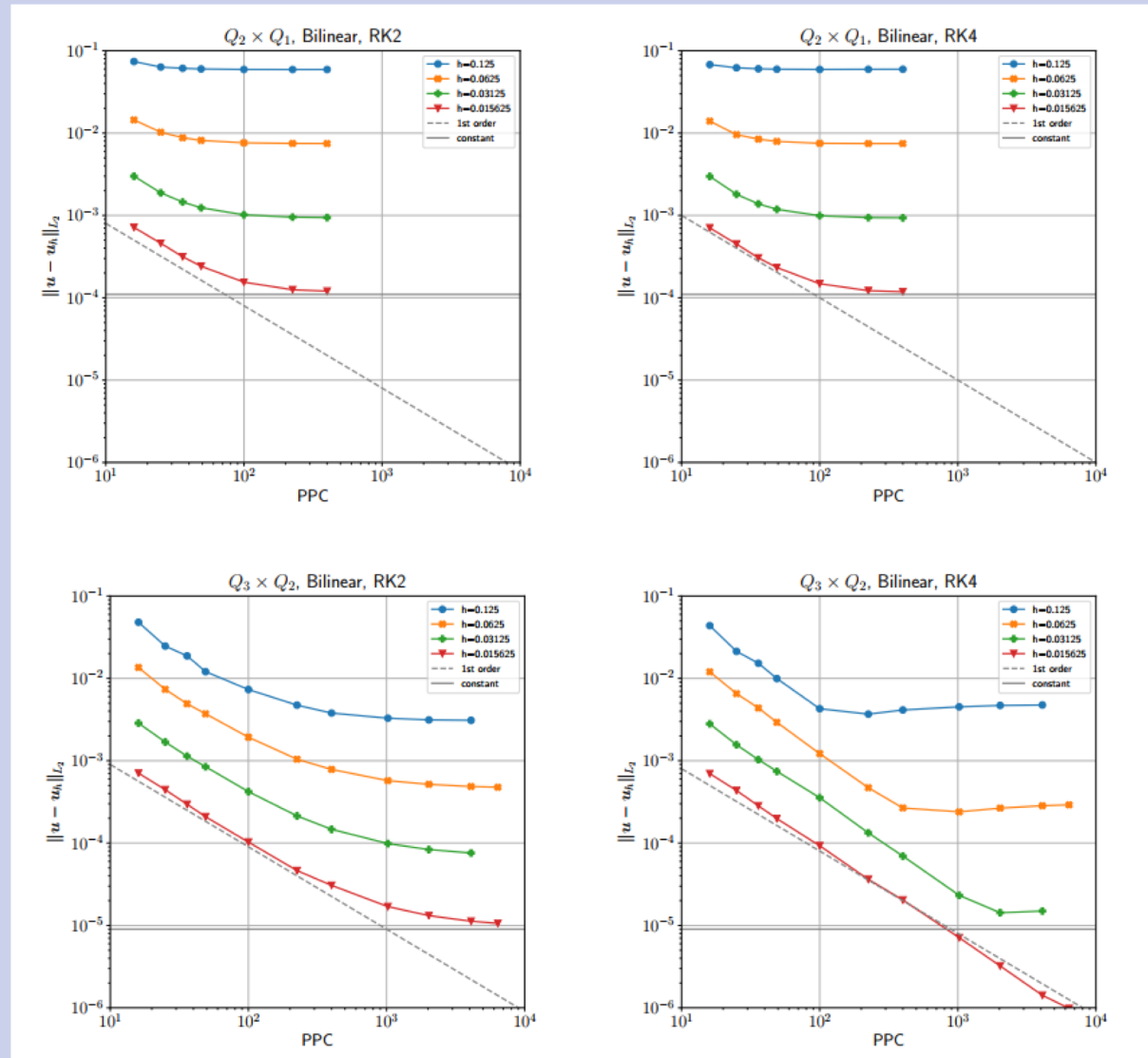
# ACCURACY: CONVERGENCE BEHAVIOR

- Particle advection scheme also limits the accuracy, but only for higher order elements
- Q2xQ1 element shows optimal convergence with RK2 integrator
- Q3xQ2 requires higher order (e.g. RK4)



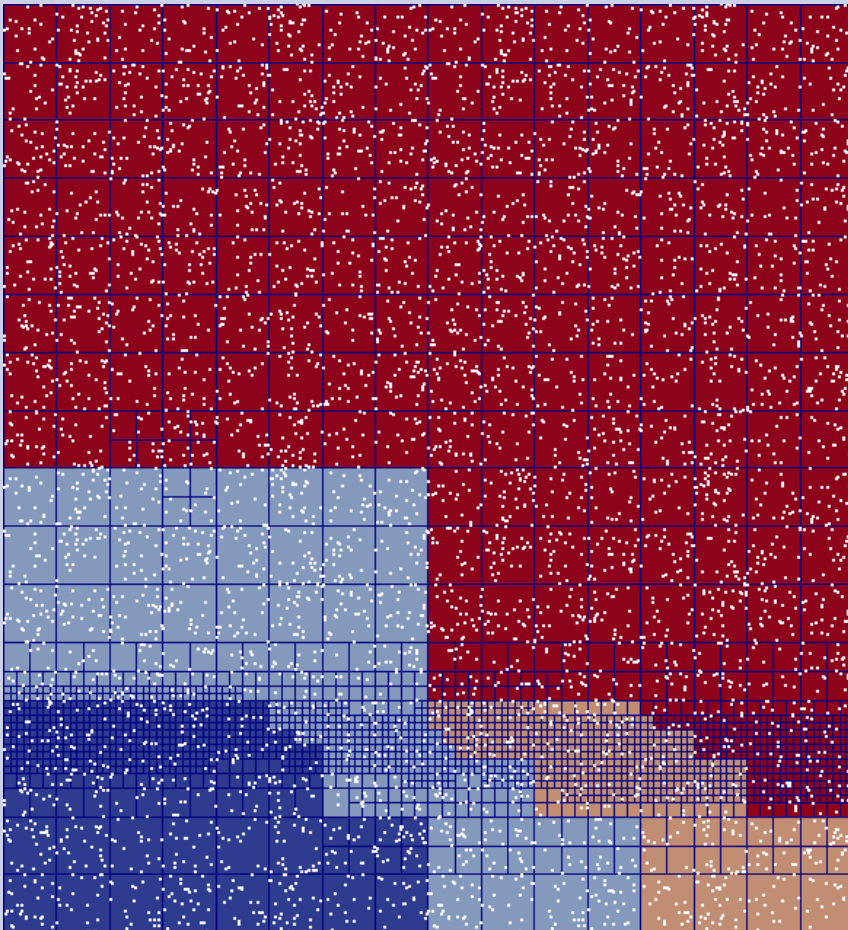
# ACCURACY: CONVERGENCE BEHAVIOR

- Results for FE order, interpolation and advection scheme shows:  
 $E(h, \text{PPC}) \sim 1/\text{PPC}$
- Thus for optimal convergence  
 (Q2Q1):  $\text{PPC} \sim h$   
 (Q3Q2):  $\text{PPC} \sim h^2$
- For typical resolutions  $\text{PPC} \leq 100$  in 2D ( $\leq 1000$  in 3D)
- Scalability?



# SCALABILITY: LOAD BALANCING

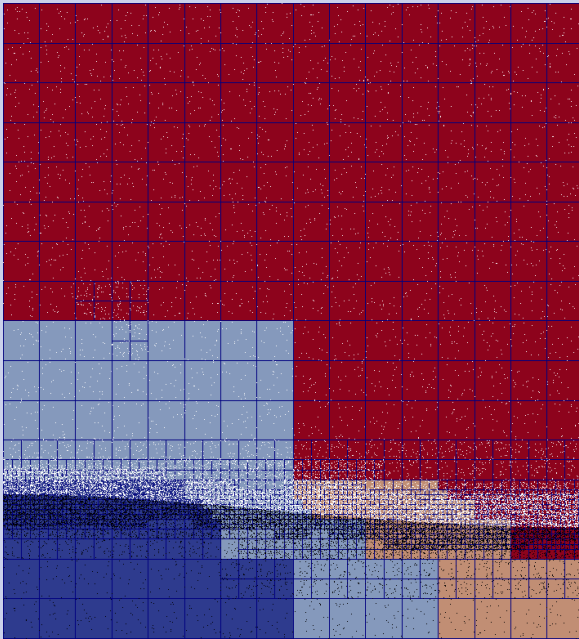
How to balance particle and cell work for adaptive meshes?



- Partitioning of domain by number of cells per process
- For uniform particle density large imbalance in particle work
- Imbalance grows with number of mesh levels
  - Limited scalability

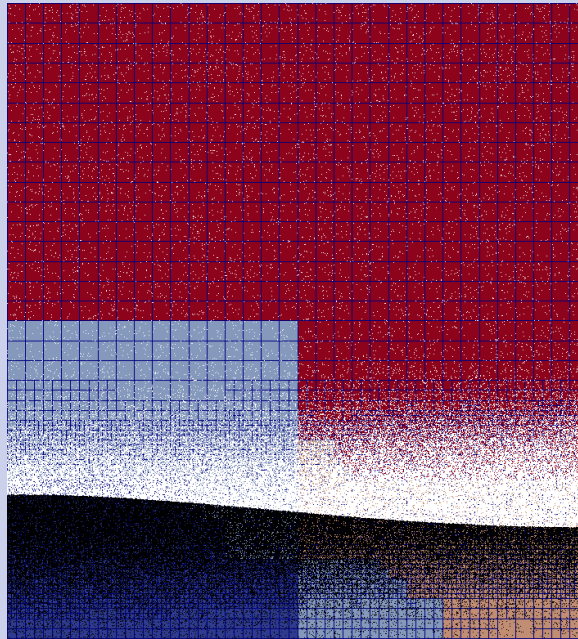
# SCALABILITY: LOAD BALANCING

Particle Management:



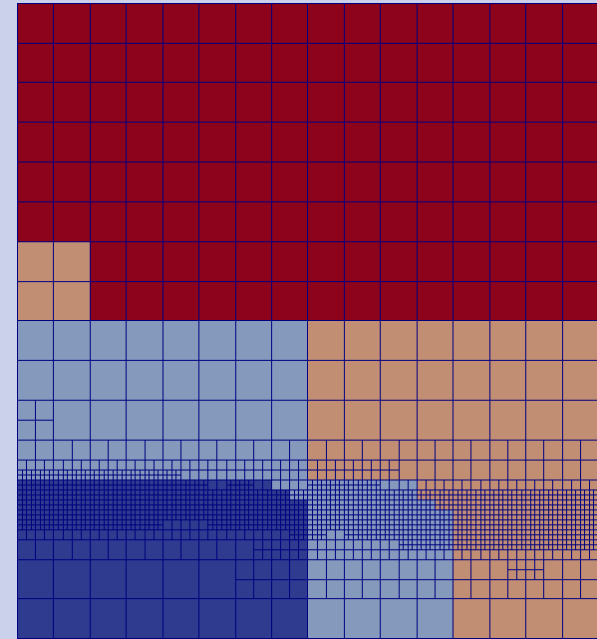
- Introduce particle population management
- Remove/add particles according to mesh
- Adds diffusion to particle properties

Variable Distribution:



- Generate variable particle distribution
- Adjust mesh according to particles
- Requires a known region of interest

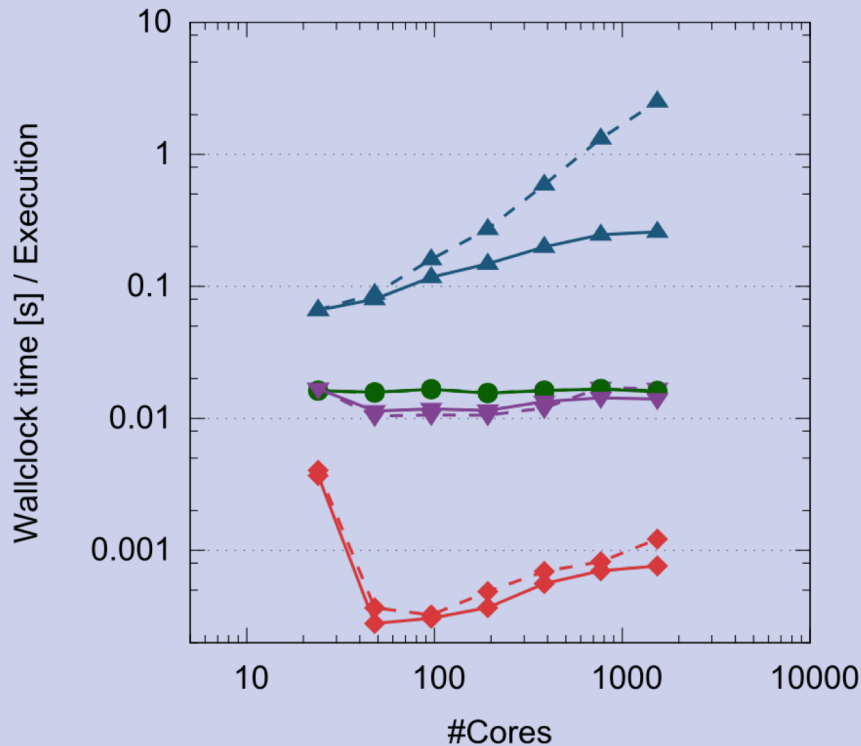
Balanced Repartition:



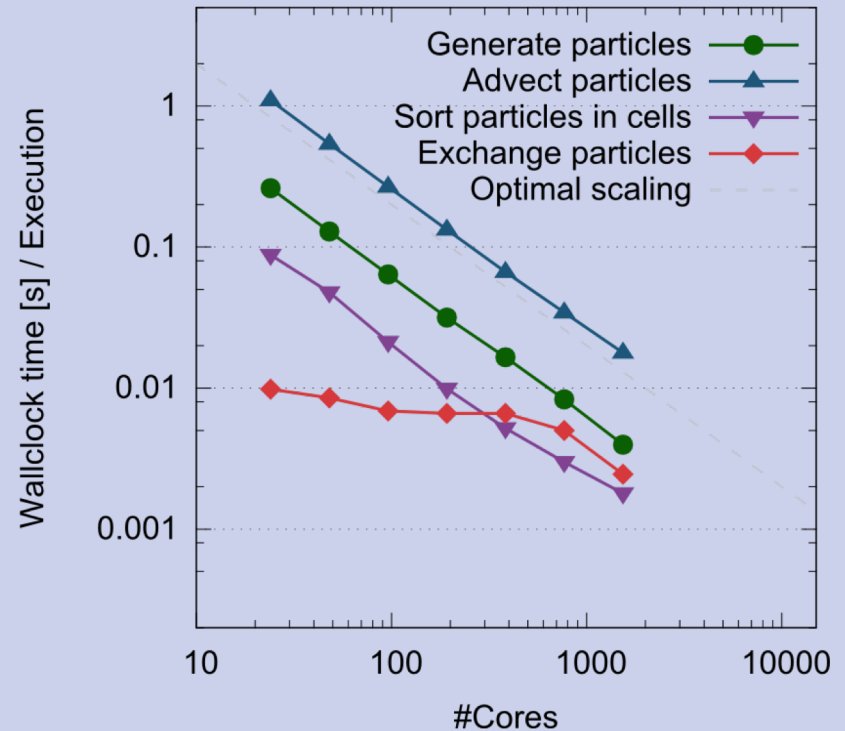
- Adjust parallel partition of mesh
- Retains identical solution
- Reaches reasonable scalability

# SCALABILITY: LOAD BALANCING

Adaptive grid - Weak scaling



Adaptive grid - Strong scaling

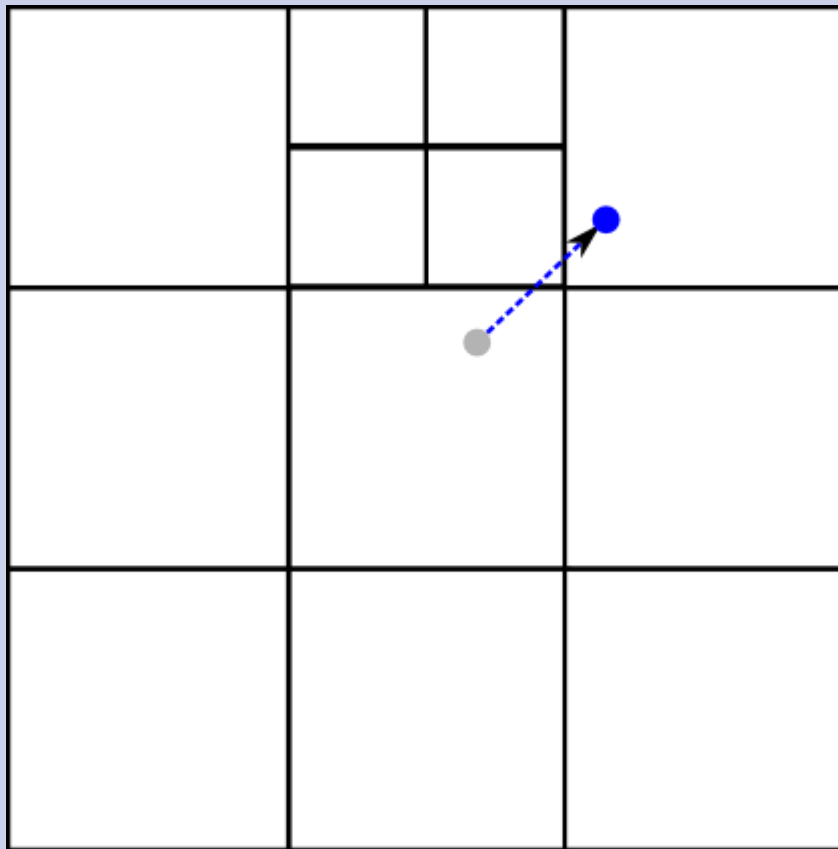


- Weak scaling dependent on load balancing technique
- Optimal strong scalability independent of technique



# EFFICIENCY: GEOMETRY INDEPENDENT

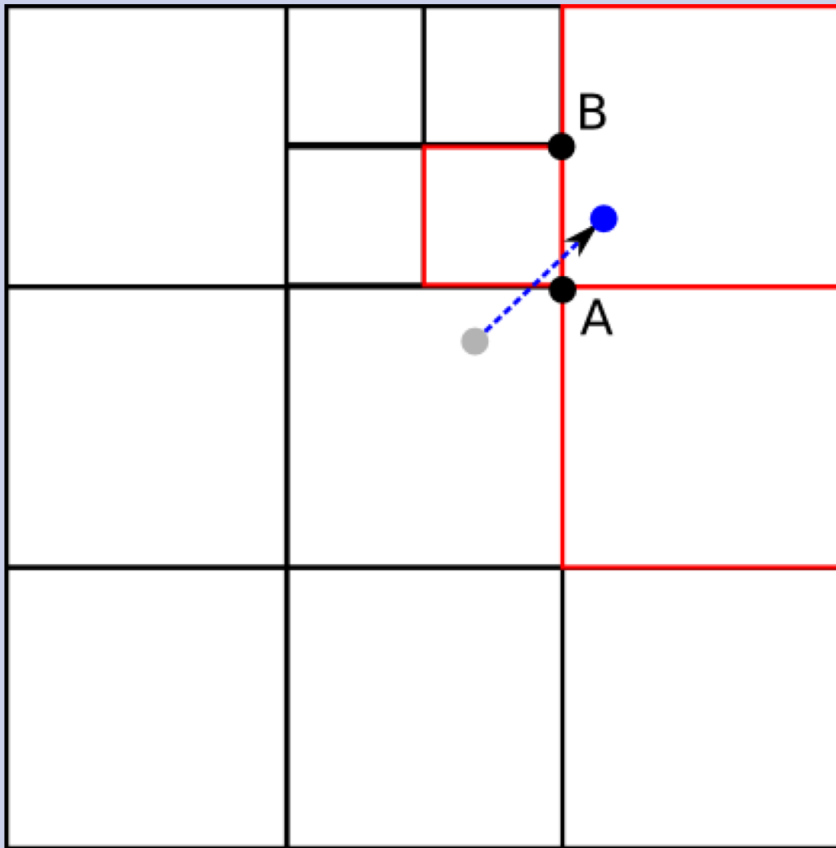
Which cell did a certain particle move to?



- Independent of geometry
- Dynamically changing mesh
- Only assumptions:
  - Quadrilateral cells
  - Hierarchical refinement
  - CFL timestep

# EFFICIENCY: GEOMETRY INDEPENDENT

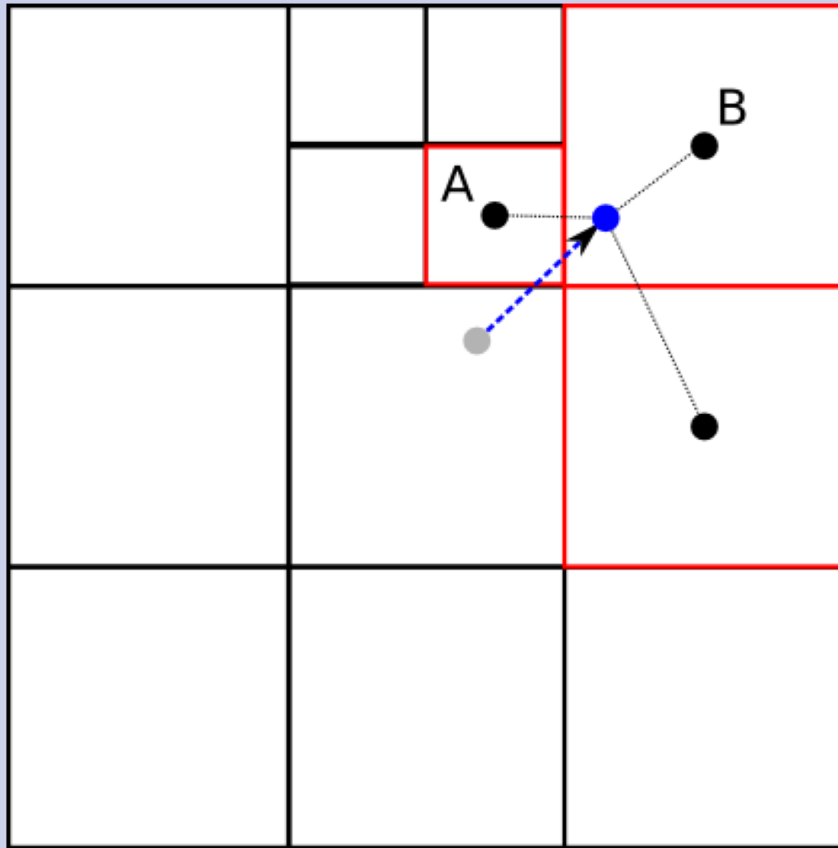
Which cell did a certain particle move to?



- Cell checks are expensive
- Check neighbors of old cell
- Check closest vertex
- Can we reduce the number of cell checks? Sort the neighbor cells?

# EFFICIENCY: GEOMETRY INDEPENDENT

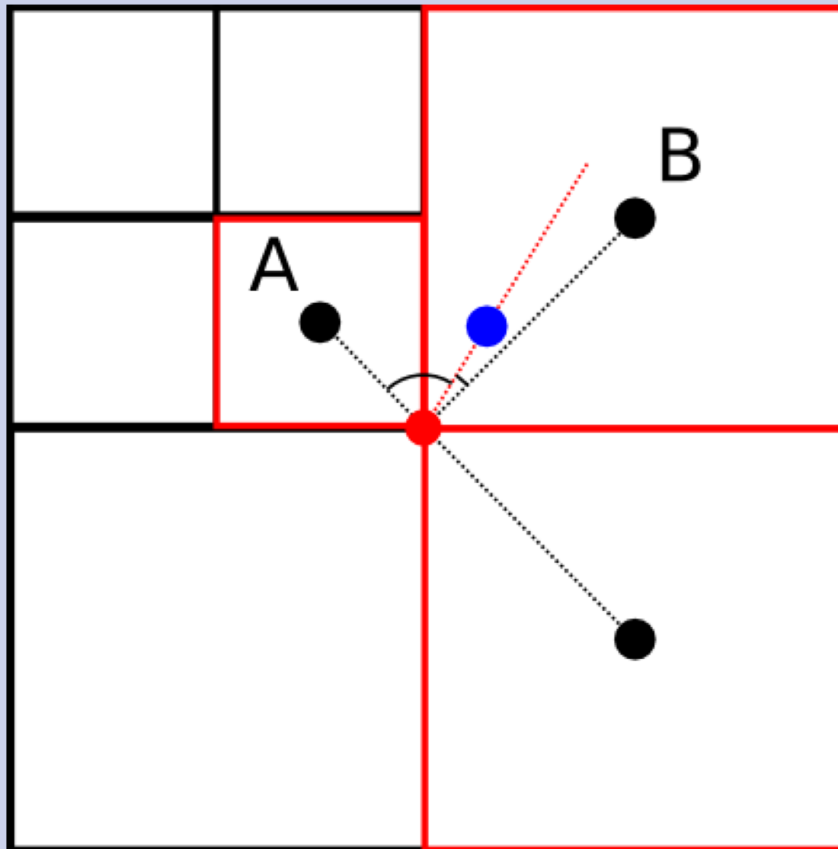
Which cell did a certain particle move to?



- Sort cells by distance to particle
- Find particle in first try for many cases
- Unreliable in adaptive meshes

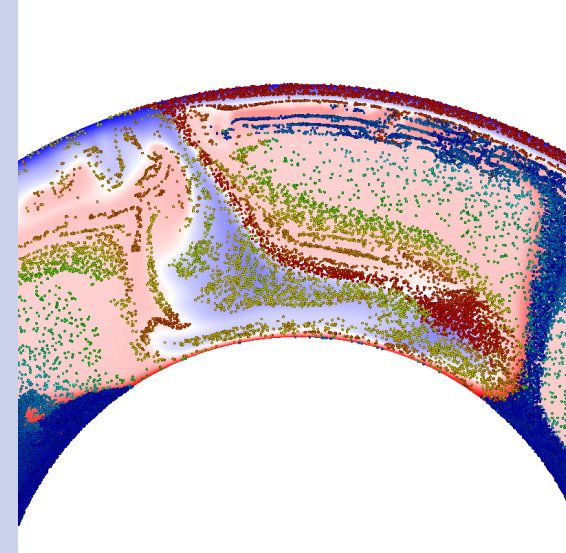
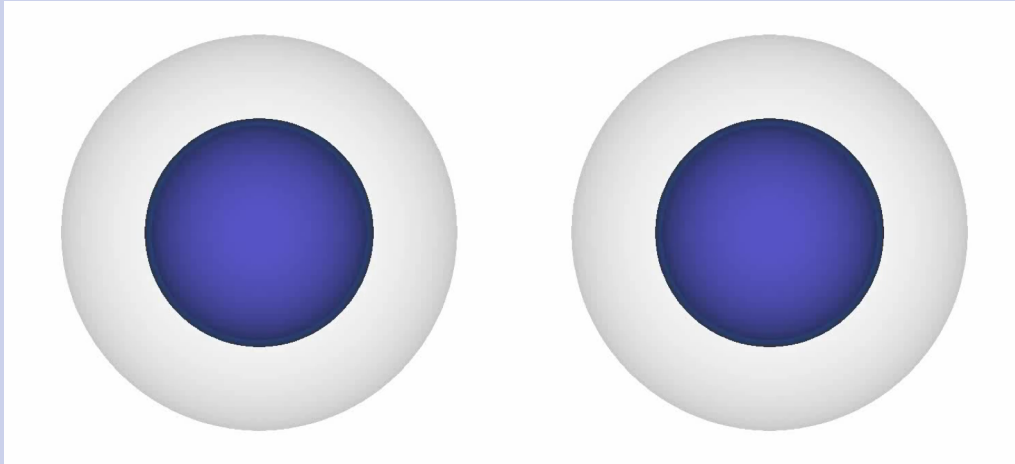
# EFFICIENCY: GEOMETRY INDEPENDENT

Which cell did a certain particle move to?

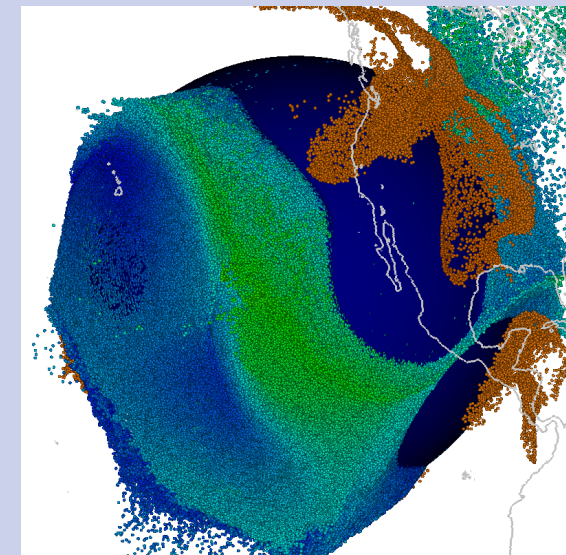


- Sort cells by angle between vertex–particle and vertex–center
- Find particle in first try for (nearly) all cases
- Reduce work by a factor of 10 compared to checking all neighbors
- Independent of geometry and mesh adaptivity

# APPLICATIONS



- Large scale mantle convection:
  - Track deformation of material
  - Track origin of material
  - Track composition of material



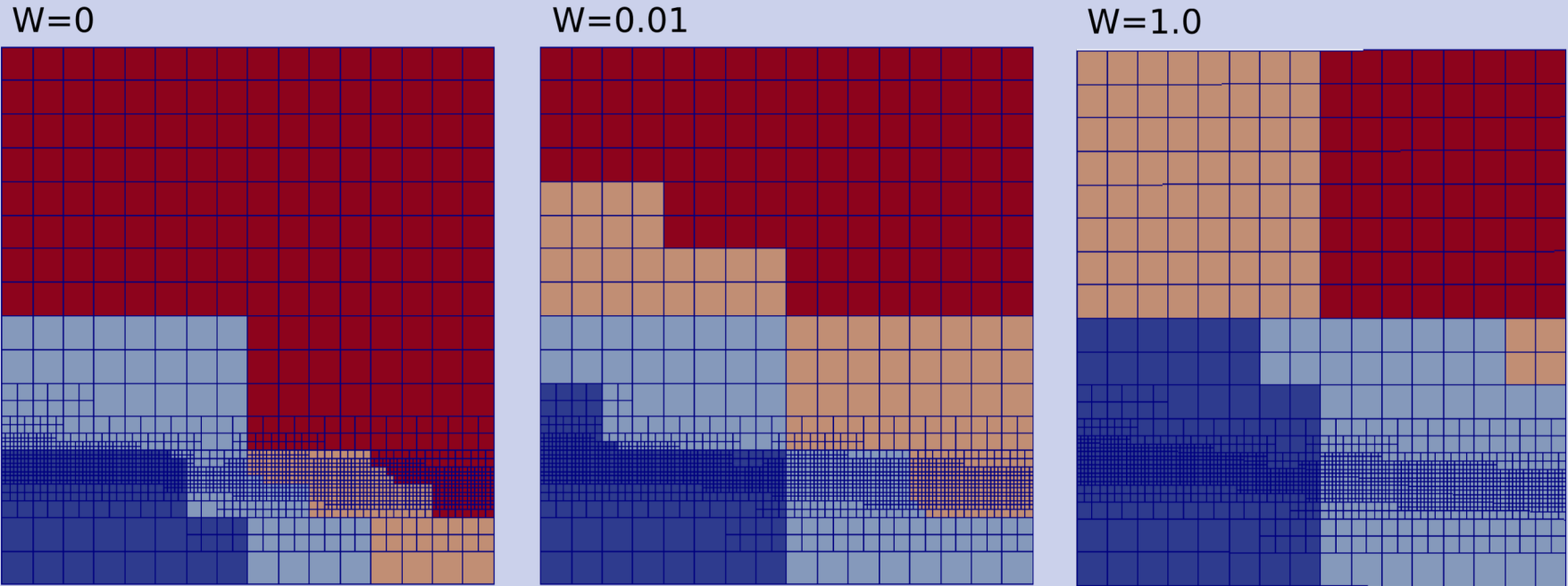
# CONCLUSIONS

- We present hybrid particle-mesh methods for use in arbitrary geometries and adaptively refined meshes
- *Convergence rate* of hybrid PIC-FE methods depends on FE method, interpolation scheme and PPC. PPC needs to increase with mesh resolution to reach higher order accuracy (not scalable).
- *Balanced repartition* load balancing achieves reasonable weak scalability without affecting the solution up to thousands of processes.
- *Angle minimization* sorting reaches optimal complexity in arbitrary geometries.

## References:

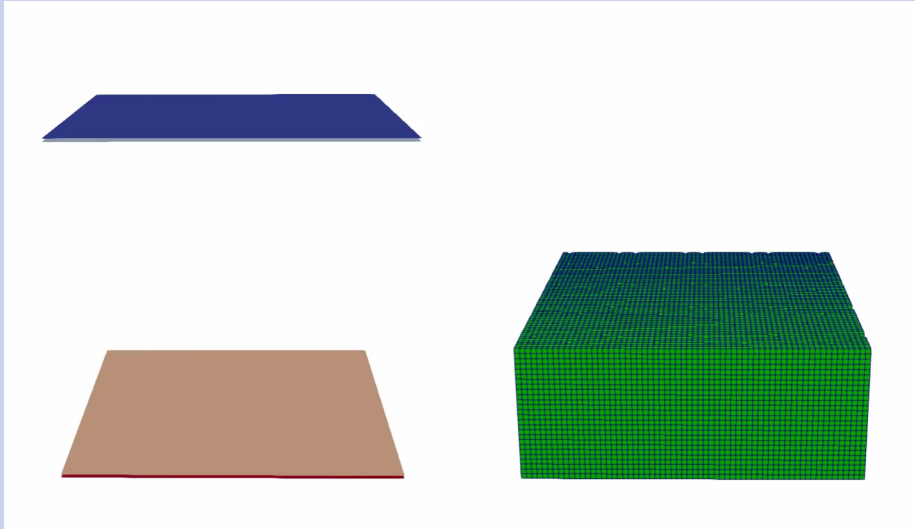
- Gassmöller, et al. "Flexible and Scalable Particle-in-Cell Methods With Adaptive Mesh Refinement for Geodynamic Computations." *Geochem. Geophys. Geosys.* 19.9 (2018): 3596-3604.
- Gassmöller, et al. "Evaluating the Accuracy of Hybrid Finite Element/Particle-In-Cell Methods for Modeling Incompressible Stokes Flow" 2019, submitted.
- *Code and benchmarks:* <https://github.com/geodynamics/aspect>

# 3. SCALABILITY: LOAD BALANCING



- Balance an appropriate sum of cells and particles
- Weight  $W$  determines the importance to balance particles
- Optimal  $W$  depends on the particle work
- Increased balancing of particles decreases balancing of cells

# APPLICATIONS



After Tackley & King, 2003

- Entrainment benchmarks:
  - Compare field methods and particles
  - Measure entrainment and convergence of PIC methods

