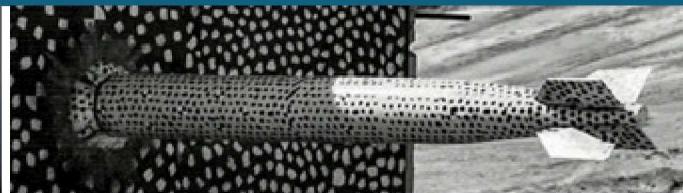


# LDRD Directors Challenge Review

An Adaptive and Scalable Meshfree Framework for  
Extreme Deformation and Failure in Solids



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FY20



Provide solid mechanics analysts with a Lagrangian method for simulating problems with very large deformation and complex multi-physics

- High velocity impact ( $\sim 100 - 3000$  m/s)
- Impact into soft targets
- Localized melting coupled with mechanical insult

#### Eulerian hydrocodes

Pros: Good robustness

Cons: Smearing of history variables.  
Difficult to model surface physics (e.g. contact, FSI)

#### Lagrangian finite element methods

Preserve material point identities  
Unambiguous surface definition

Lack of robustness due to mesh tangling; Reconnection helps (e.g. LGR)

#### Lagrangian mesh-free methods

Better smoothness than finite elements.  
Reconnection easier with no element topology constraints.

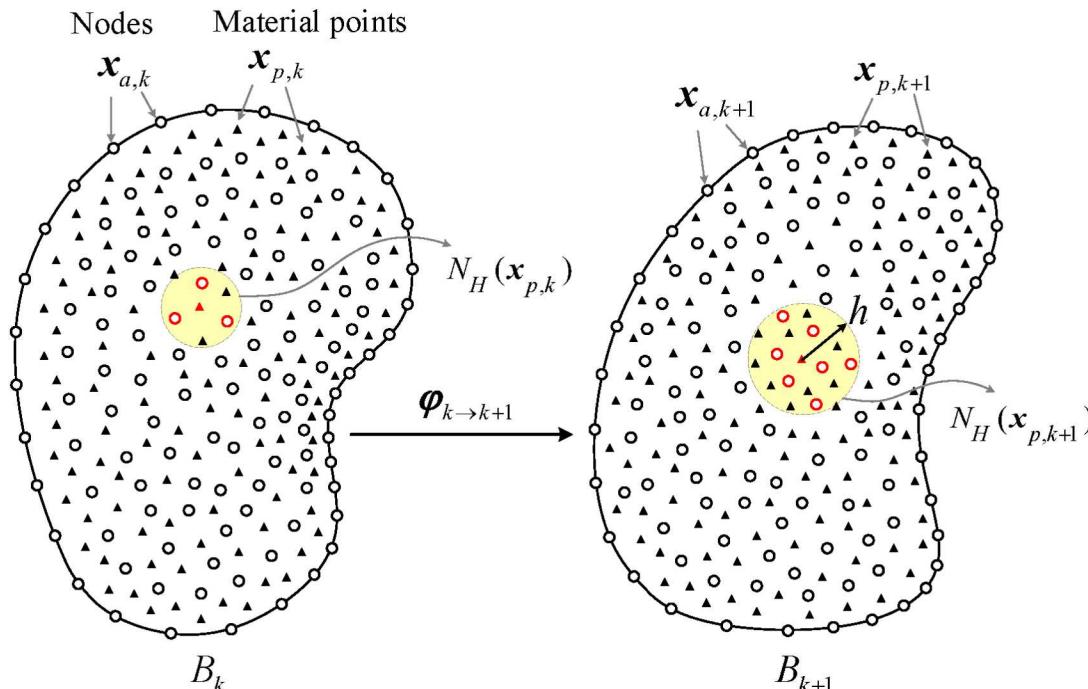
Fixed node & material point set may be insufficient to guarantee accuracy and robustness

## Technical approach

New mesh-free adaptivity scheme: **node/point injection within the Optimal Transportation Meshfree (OTM) method**

OTM<sup>1</sup> provides

- Geometrically **exact** updated Lagrangian kinematics
- Exact **conservation** of linear momentum and angular momentum
- **Compatibility** with many mesh-free interpolation schemes, including RKPM. We used local max-ent.
- A variational **formalism**, shared with finite element methods, that provides strong mathematical guarantees.



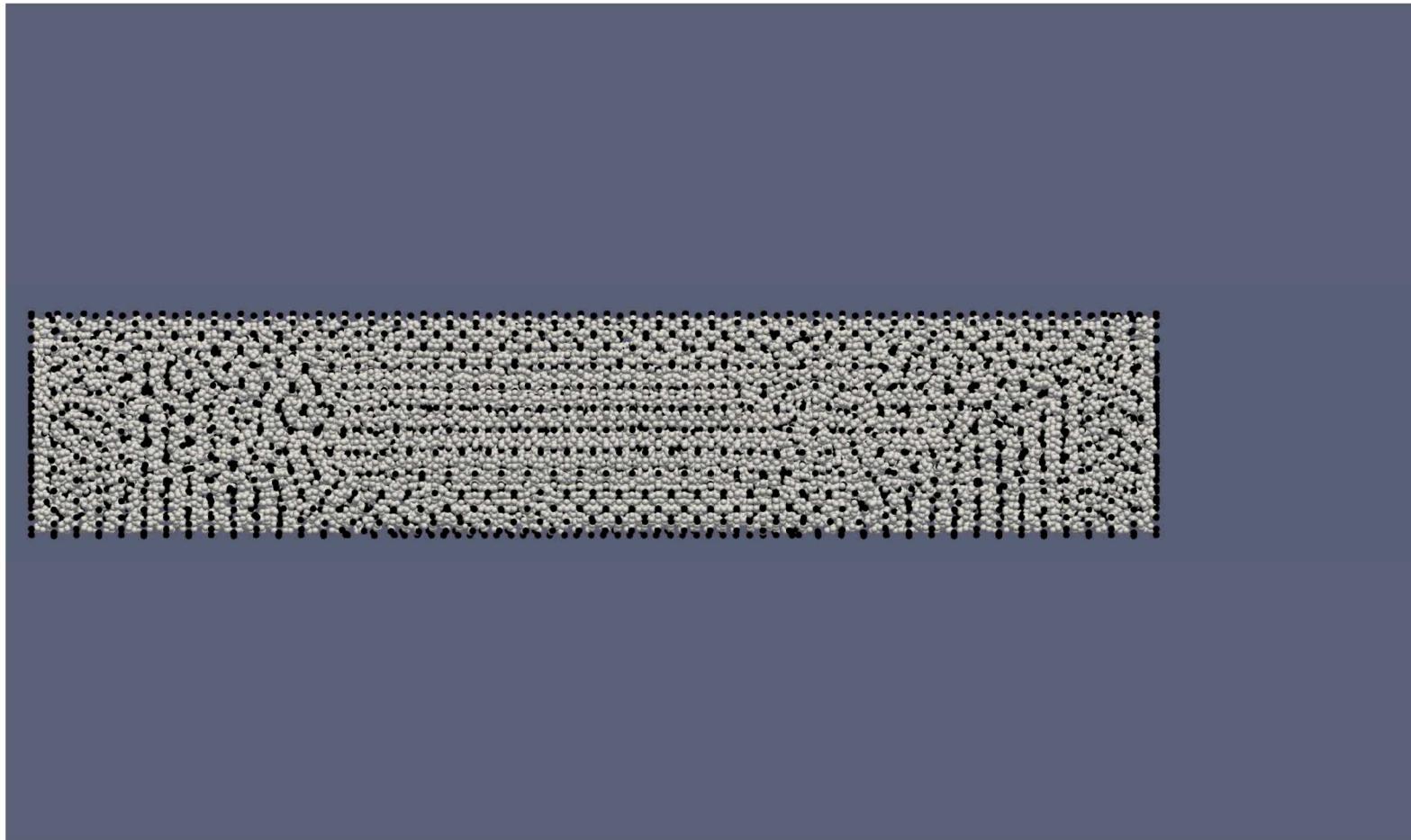
Material (quadrature) points are fully Lagrangian

**Get reconnection without mapping**

$$x_{p,k+1} = \varphi_{k \rightarrow k+1}^h (x_{p,k})$$

$$\varphi_{k \rightarrow k+1}^h (x) = \sum_a x_{a,k+1} N_{a,k+1}$$

# OTM: The meshfree Optimal Transportation Method



Like FEM, OTM has:

● Nodes

○ Material Points

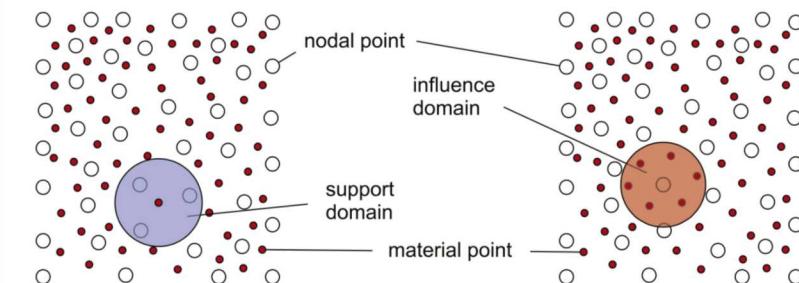


Unlike FEM, OTM has no:

■ Mesh or Elements

**Points have support nodes**

**Nodes have influence points**



Weißenfels and Wriggers, CMAME, 2017



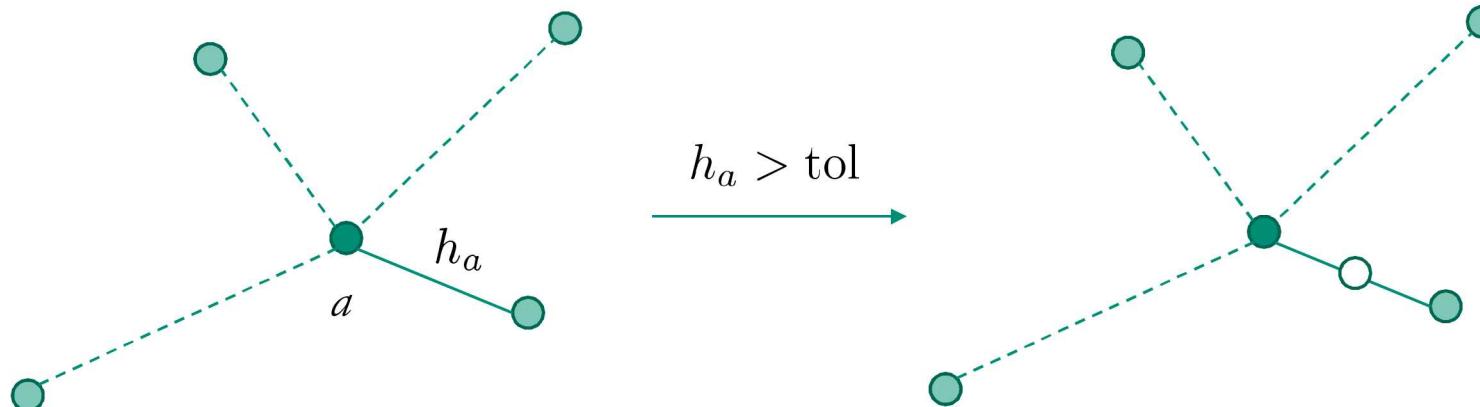
## OTM node and point injection

Introduce new nodes and material points on the fly where needed

Existing nodes and material points remain. **No mapping of state variables.**

Do need to interpolate state variables to new material points – we use Lie group interpolation and variational recovery to ensure they stay in their proper manifolds<sup>1</sup>

Initial focus is on capability demonstration. We set a very simple adaptivity metric:



```

for  $a \in \text{nodes}$  do
   $h_a \leftarrow \min_{b \neq a} |x_a - x_b|$ 
  if  $h_a > \text{tol}$  then
    insert node halfway between
    interpolate fields to new node
  end if
end for

```

- Similar scheme for nodes and material points
- Executed periodically (every N timesteps, say)

<sup>1</sup> Mota et al. *Comp Mech* 2013, 52

## Future work on adaptivity metrics

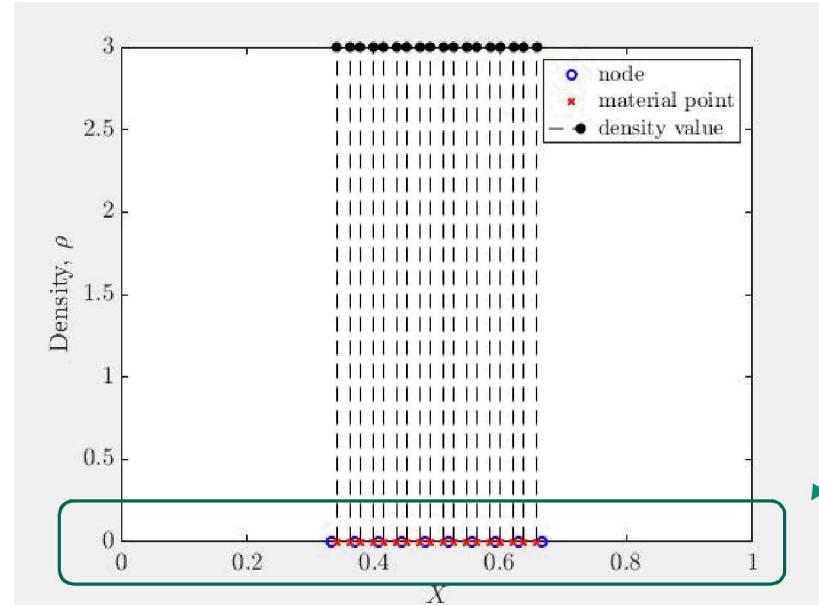
### Variational node/point injection

OTM is based on **minimization** of an effective energy functional (true even with history dependence)

Introducing new nodes changes this energy, depending on position. Similar to diffusion of species.

Explore using this analogy to optimize new nodal locations (and choose optimal number of nodes given a cost). Use **physics** of problem to **drive adaptivity** directly.

Use recent OTM treatment of diffusion<sup>1</sup> to drive nodes to **optimal locations** during adaptation step



- Nodes and points initially clustered in center of domain
- Distribute themselves uniformly due to entropy of mixing
- Combining with mechanical energy functional would bias them towards under-resolved features

# LGR: a performance-portable Lagrangian remeshing code toolkit



## LGR: Toolkit for Lagrangian Grid Reconnection, Dan Ibañez 1443

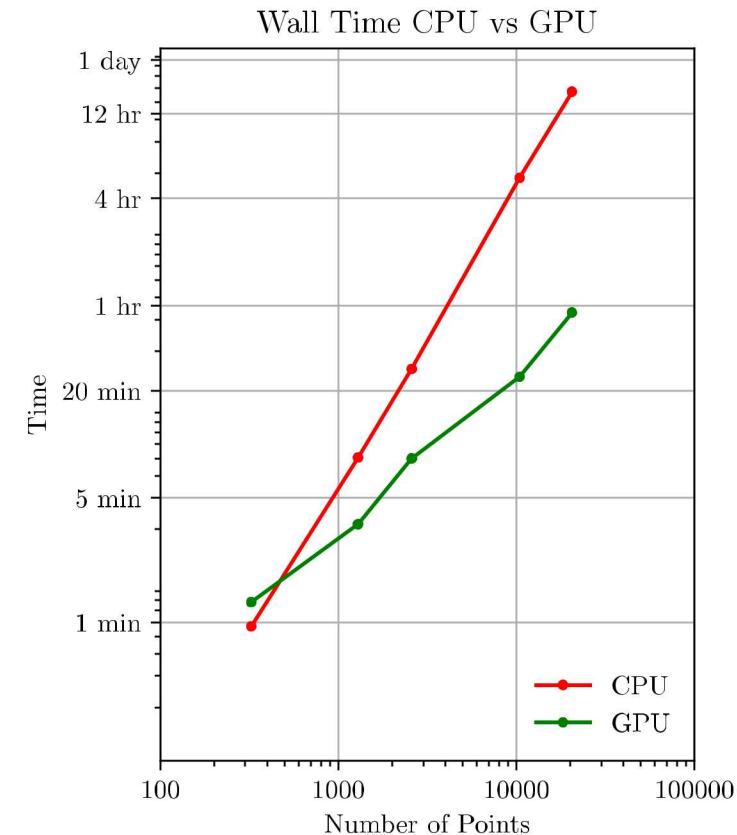
- **Open-source** on Github
- Fundamentally designed from the ground up for **adaptivity**, a key component of our project
- Has minimal dependencies, enabling **rapid prototyping of ideas** and quick build-test turnaround times

## LGR on NVIDIA GPUs

- **Performance portability** based on raw CUDA code, Thrust library, and C++14 **standard** constructs
- **Interoperable** with other CUDA-enabled libraries (Kokkos, ArborX...)
- Data access patterns designed for **fast access** and **minimal data race conditions** on threaded architectures

## Our team

- Gained valuable experience **developing portable algorithms and testing complex code** for correctness and performance on both the CPU and GPU **without any algorithm duplication**
- Found that OTM methodologies are amenable to **performance on GPU architectures**—further exploration and optimization required!
- Utilization of GPU for our code is **above 90%** most of the time.





## Geometric search

- is at the core of meshfree methods to establish node/point relationships (akin to connectivities in FEM)
- has broader applicability, e.g., in **contact algorithms**

**ArborX:** an open-source library designed to provide performance portable algorithms for geometric search

- Hosted on Github
- Based on Kokkos and Thrust libraries to deliver performance on both CPU and GPU architectures
- Currently under investigation in other Sandia apps (ATDM and IC: Sierra, NimbleSM...)

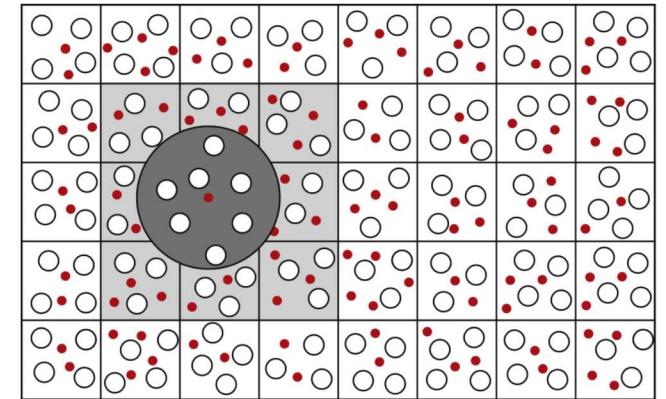


Fig. 4. Subdivision of the domain into static cells.

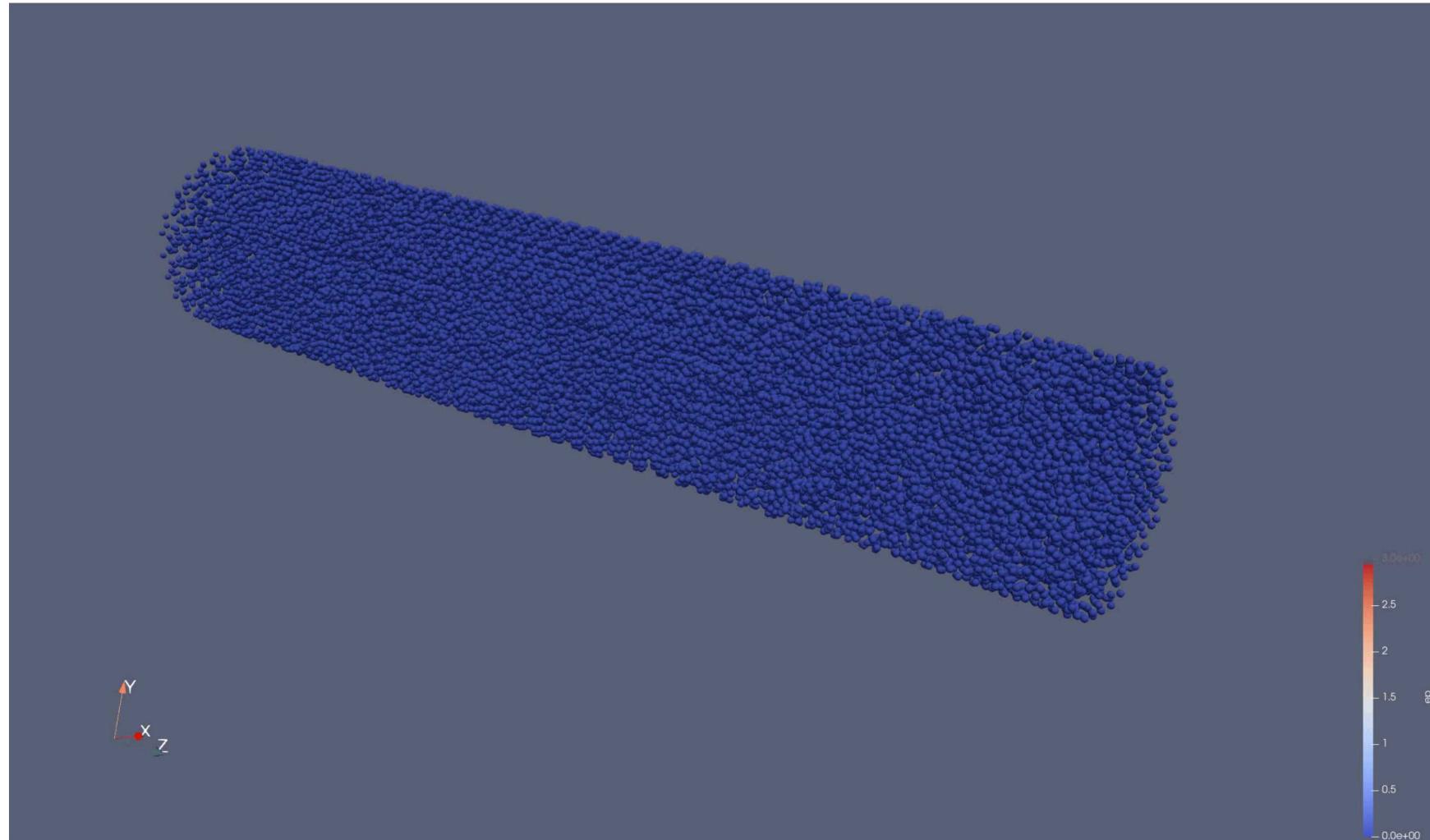
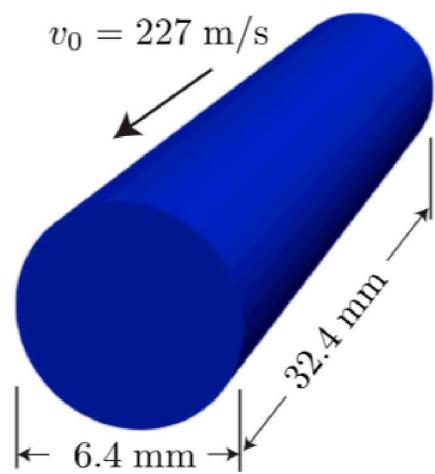


[github.com/arborx/ArborX](https://github.com/arborx/ArborX)

We developed a **reusable, performance-portable search library interface** for LGR

## Taylor Bar with OTM running on LGR v3 on the GPU

Copper Taylor bar of length 32.4 mm and diameter 6.4 mm impacting a rigid wall with an initial velocity of 227 m/s. Common example for testing numerical methods for large deformation (Li, Habbal & Ortiz 2010).



# Taylor Bar with OTM running on LGR v3 on the GPU

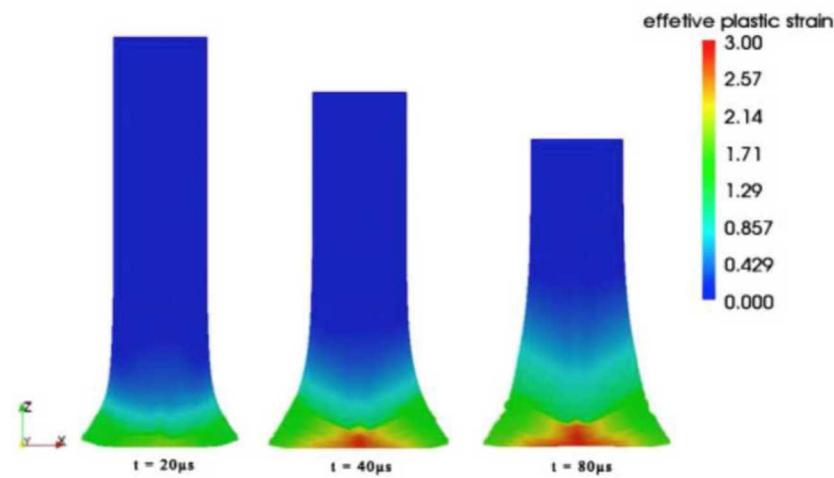
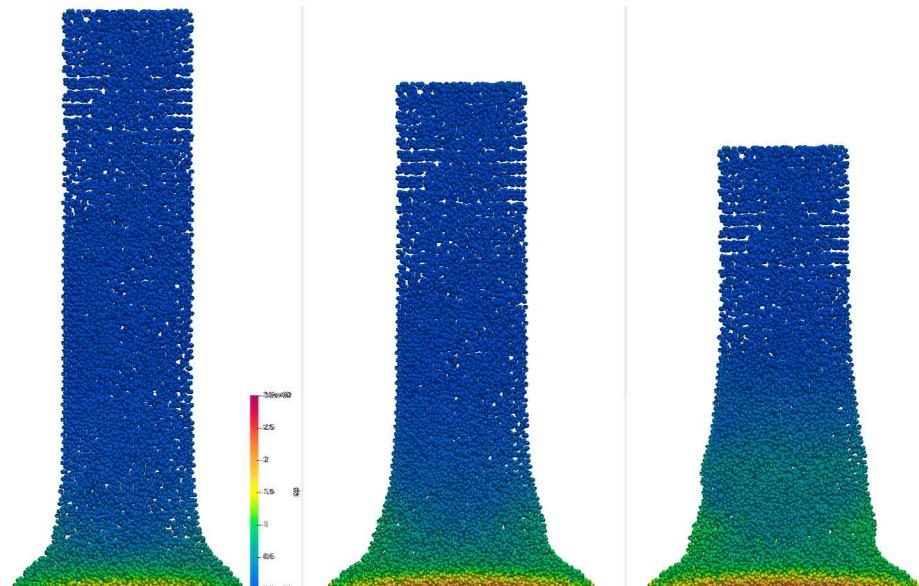
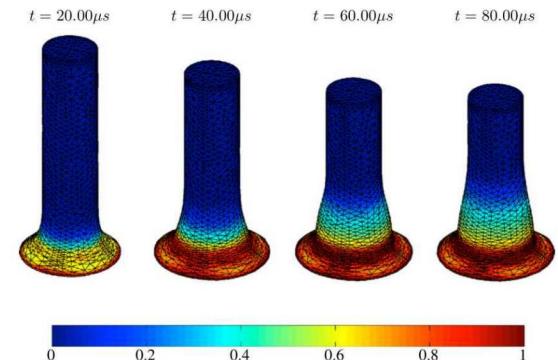


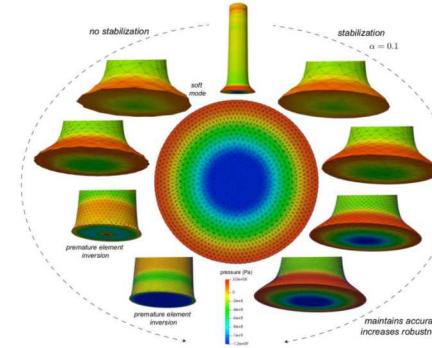
Figure 14. Taylor-anvil impact test of copper specimen at 227 m/s impact velocity. Distributions of effective plastic strain at 20, 40 and 80  $\mu$ s.



Li, Habbal & Ortiz (2010) eqps results  
 $t = 20\mu\text{s}, 40\mu\text{s}, 80\mu\text{s}$   
 5174 nodes, 26741 points  
 They look very smooth.  
 Postprocessed from point data?



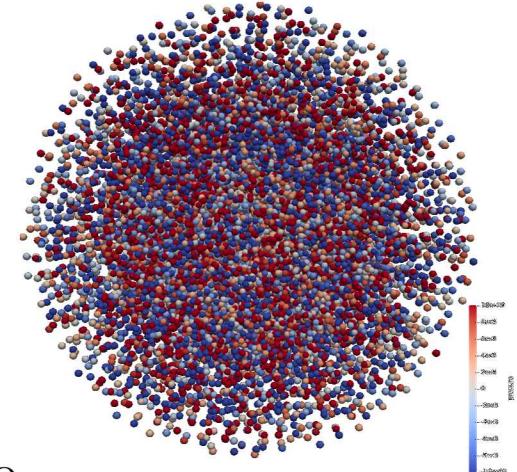
Aguirre et al., JCP, 2014



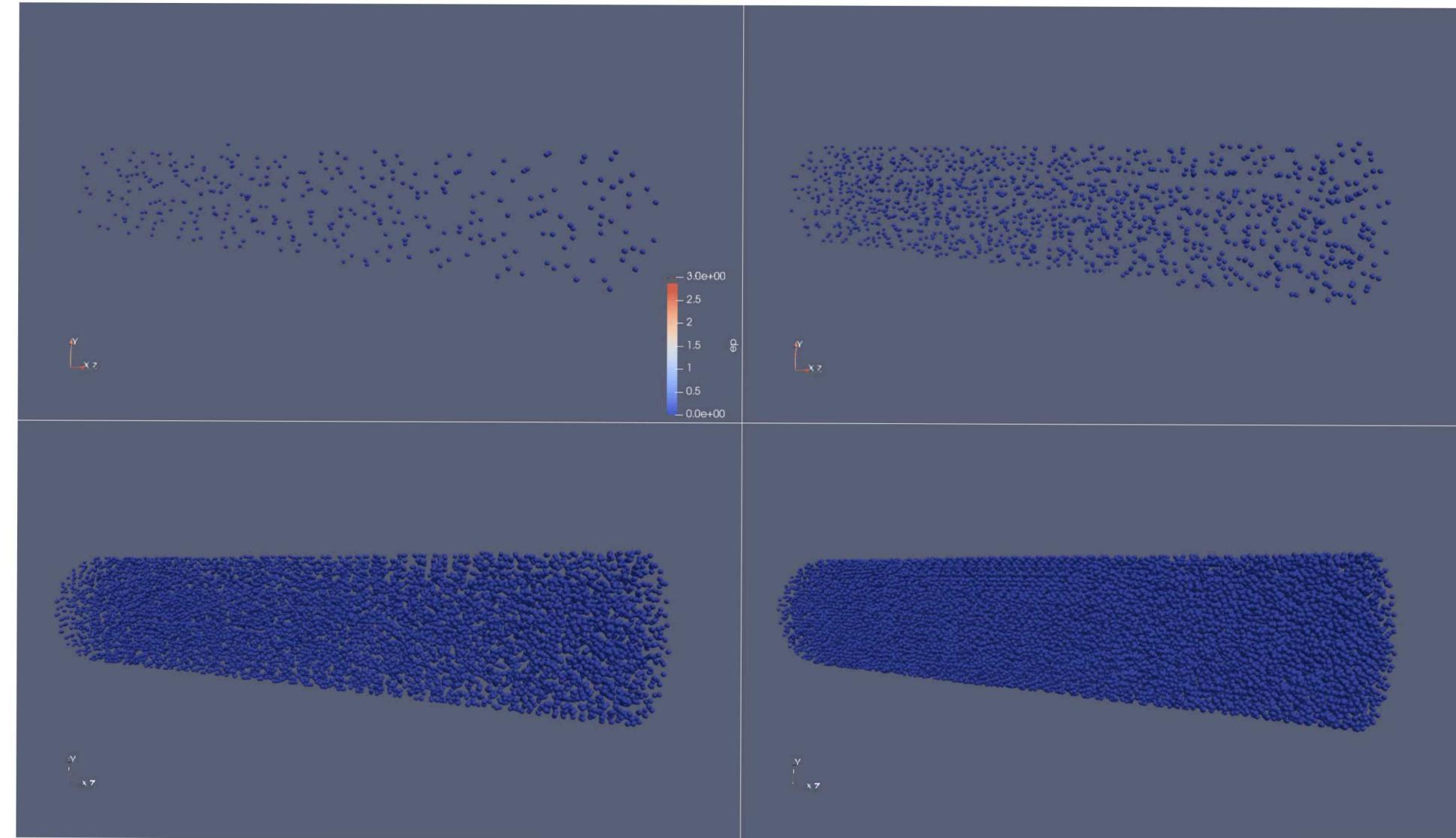
Our eqps results      Foulk et al., IJNME, 2020?

$t = 20\mu\text{s}, 40\mu\text{s}, 80\mu\text{s}$   
 5179 nodes, 26488 points

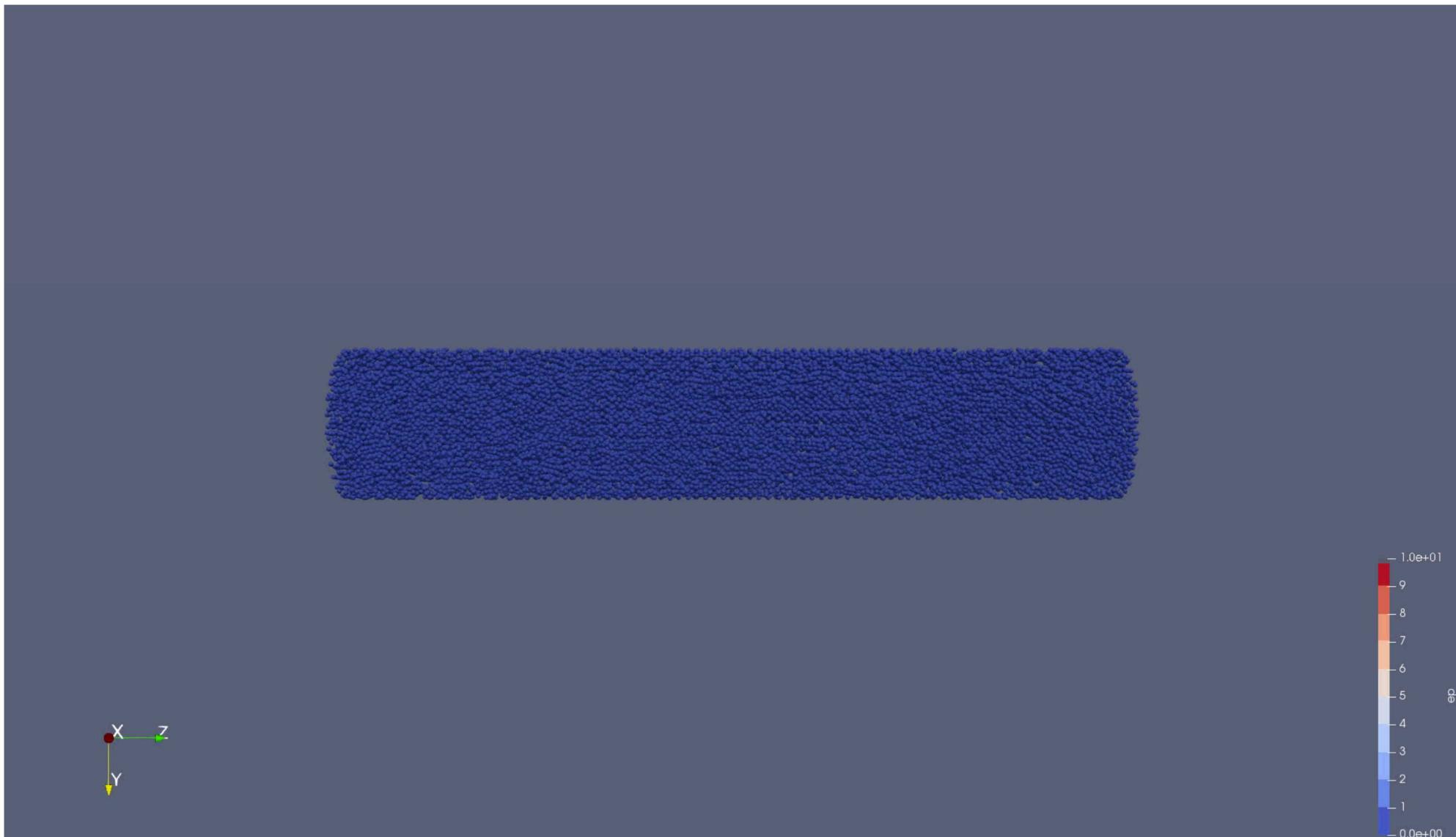
These are raw point results without postprocessing.  
 They are not perfectly symmetric because we initialized with an unstructured mesh generated by Cubit.  
 The view is from a slice down the middle.  
 Pressure oscillations due to locking. Mitigate (Tupek & Koester).



# Taylor Bar with meshfree OTM and adaptation



# Taylor Bar at Extreme Deformation



## Key Accomplishments

- Developed and implemented a framework for **meshfree** simulation.
- Leveraged the theory of the Optimal Transportation Method for **extreme deformation**.
- Introduced **adaptivity** to meshfree simulations.
- Leveraged the LGR code for **scalability** and execution on the GPU.
- Our team achieved the first year objectives.

## Lessons Learned

- Holistic synergy between 1400, 1500, and 8300.
- Strong, extremely capable team.
- Can achieve regimes of extreme deformation naturally.
- Can introduce adaptivity with relative ease.
- OTM inherits many fundamentals from FEM.
- Created a tool for further exploration of meshfree methods, adaptivity, and extreme deformation.

## Risks, Issues, and Mitigations

- Old habits are hard to break (no mesh, no elements).
- Some things look hard but are easy and vice versa.
- Develop new intuition through experience.
- Better adaptivity metrics (variational, energy based).
- Shape functions and adaptivity at extreme deformation (use of different schemes).
- Locking in isochoric deformation (Mike Tupek and Jake Koester looking at this).

## Future Promise

- Keep our team.
- Formulate novel, meshfree adaptivity metrics.
- Access regimes currently inaccessible with existing technology:
  - impact on hard, soft targets
  - simulation of subtractive and additive manufacturing
  - phase changes, melting
  - pervasive fracture and fragmentation
  - fire scenarios and others