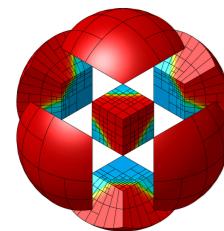


Interpolation at Arbitrary Points in High-Order Meshes on GPUs



MFEM Community Workshop

22-24 October 2024



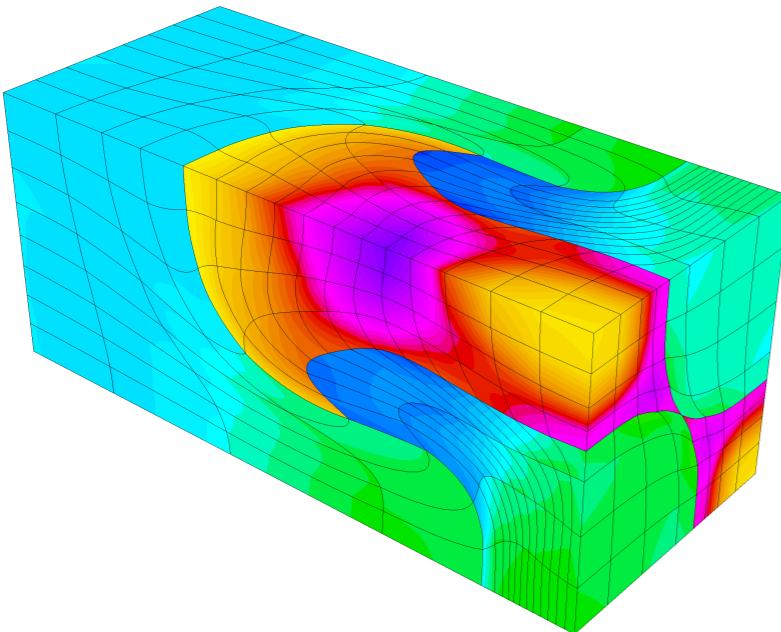
Ketan Mittal

Aditya Parik (USU), Tzanio Kolev (LLNL)

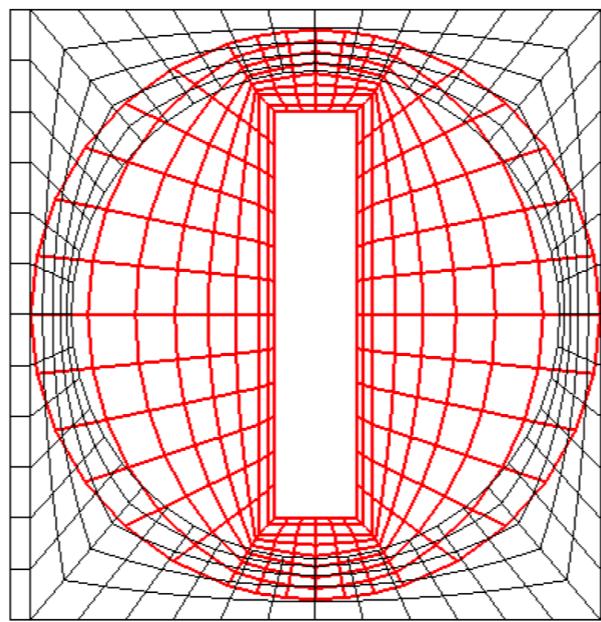


Motivation

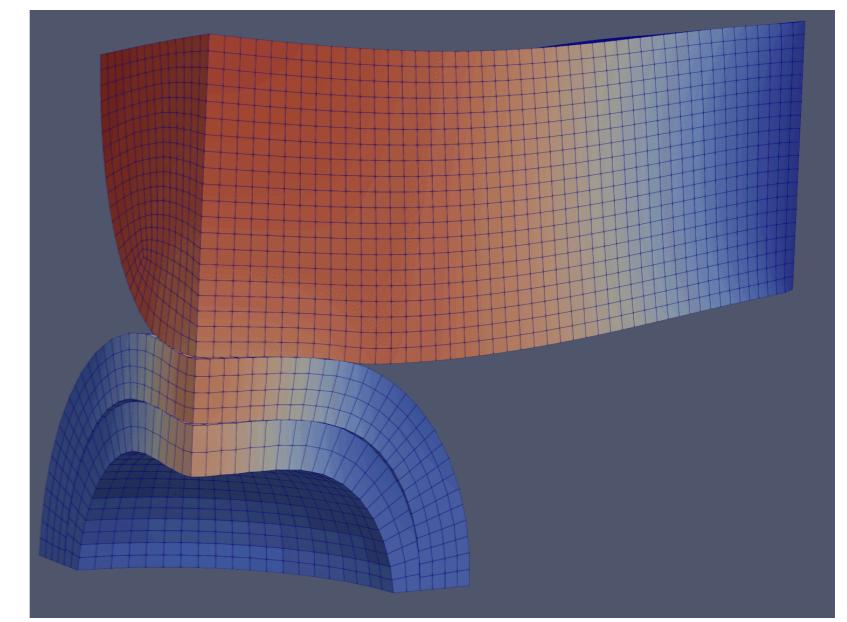
- Interpolation at arbitrary points is required in FEM for:
 - Querying the solution at desired locations.
 - Exchanging information between overlapping grids.
 - Detecting contact between meshes.



$p = 3$ mesh for triple-point problem



$p = 2$ meshes for overlapping grids



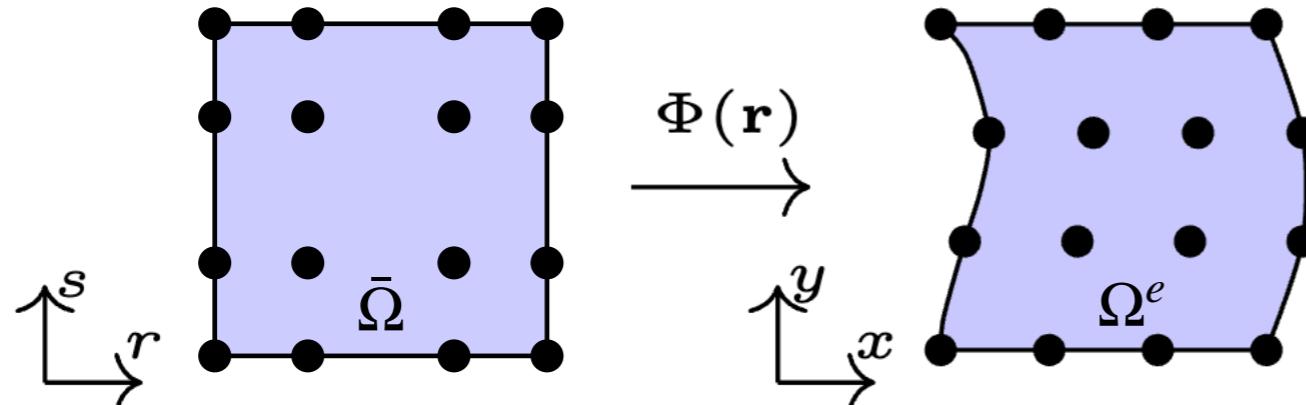
$p = 1$ meshes in contact

This is a challenging problem, especially for unstructured curvilinear meshes distributed on many MPI ranks in HPC.

Background

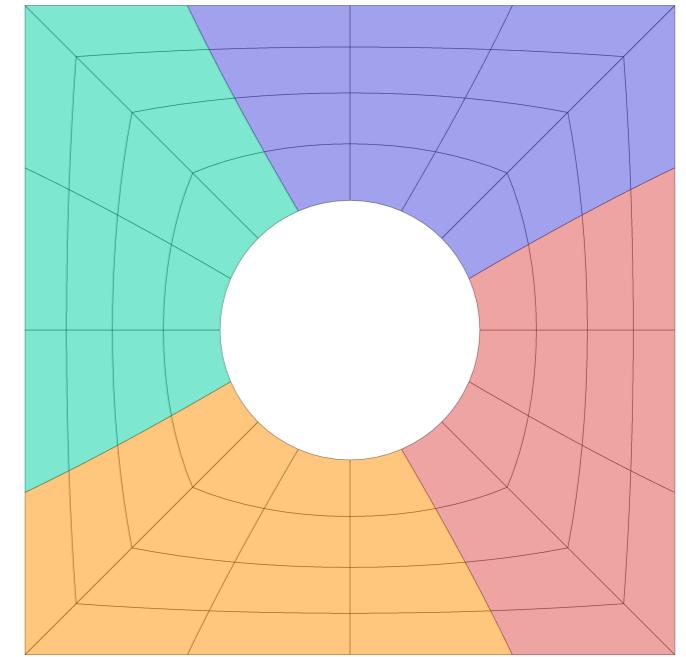
- Each *tensor-product* element in the mesh is represented using Lagrange interpolants $\phi_i(\mathbf{r}), i = 1 \dots N^D, D = [1,3]$, on Gauss-Lobatto-Legendre points in the reference element $\bar{\Omega} \in [0,1]^D$.

- $$\mathbf{x}(\mathbf{r})|_{\Omega^e} = \sum_{l=1}^{N^D} \mathbf{x}_l^e \phi_l(\mathbf{r}) = \sum_{i=1}^N \sum_{j=1}^N \sum_{k=1}^N \mathbf{x}_{ijk}^e \phi_i(r) \phi_j(s) \phi_k(t), \quad \mathbf{r} \in \bar{\Omega}$$



- Similar mapping is used for any function, e.g., Velocity $\mathbf{u}(\mathbf{r})$ and Temperature $T(\mathbf{r})$.

For a given point \mathbf{x}^* , we need to know the element e^* on MPI rank p^* that overlaps the point, and the corresponding reference-space coordinates (\mathbf{r}^*) inside Ω^{e^*} .



Quad mesh on 4 MPI ranks

Methodology

- **One time setup**

- Compute data structures to quickly map a given point, $\mathbf{x}^* = \{x^*, y^*, z^*\}$, first to MPI ranks (\underline{p}) and then to candidate elements (\underline{e}) locally on each rank.
- Compute element-wise bounding boxes to quickly test if a given point is inside an element.

- **Search a given set of points**

- Newton search to compute reference-space coordinates in the candidate elements:

$$\underset{\mathbf{r}}{\operatorname{argmin}} \frac{\|\mathbf{x}^* - \mathbf{x}(\mathbf{r})\|_2^2}{2}$$

- **Interpolate the discrete solution**

- $u\left(\mathbf{x}^* = \Phi(\mathbf{r}^*)\right) = \sum_{i=1}^N \sum_{j=1}^N \sum_{k=1}^N u_{ijk}^e \phi_i(r^*) \phi_j(s^*) \phi_k(t^*)$

*Implementation based on **gslib** developed by James Lottes in the context of SEM for Nek5000*

Setup



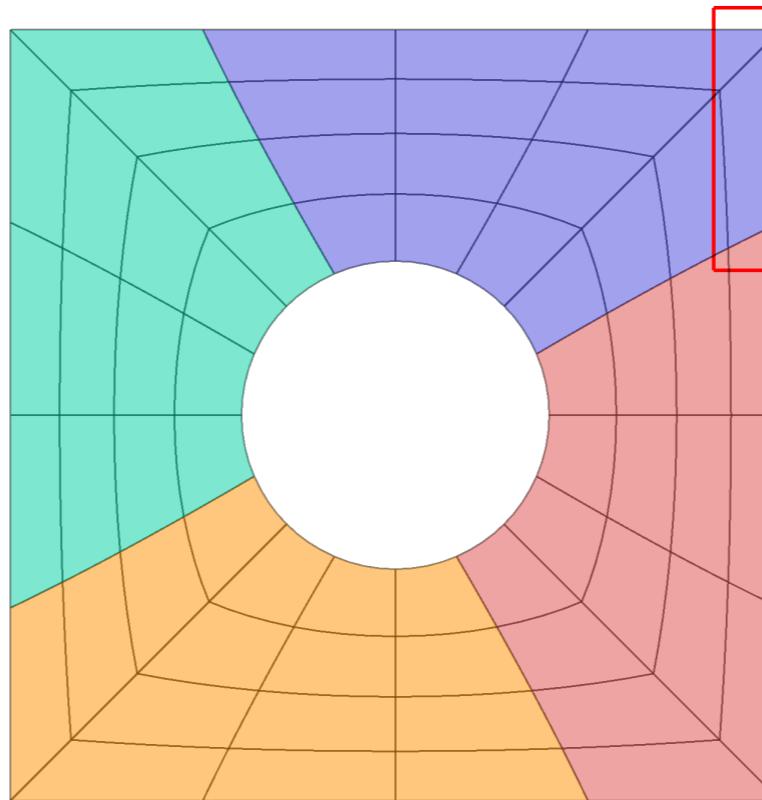
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Axis-Aligned Bounding Box

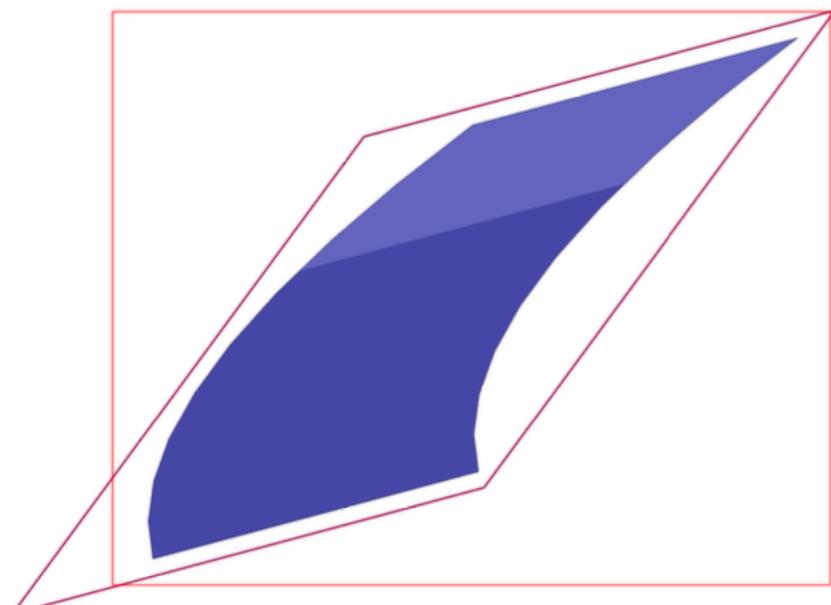
- Compute bounds on $\mathbf{x}(\mathbf{r})|_{\Omega^e} = \sum_{i=1}^N \sum_{j=1}^N \mathbf{x}_{ij}^e \phi_i(r) \phi_j(s)$ to determine $\{x_{\min}, x_{\max}, y_{\min}, y_{\max}\}$.



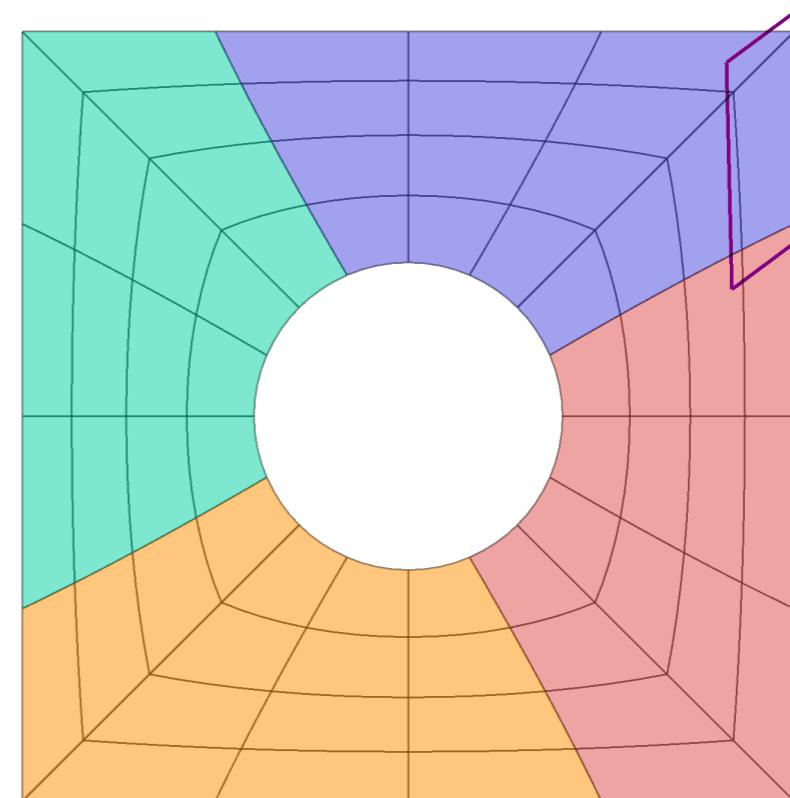
Axis-Aligned Bounding Boxes

Oriented Bounding Box

- AABB are suboptimal for curvilinear elements.
- OBB provide tighter bounds around each element.
 - Represented by OBB center and a transformation matrix ($A_{D \times D}$) with respect to the reference element.



Bounding boxes around a curvilinear element



Oriented Bounding Boxes

Local Map for $\mathbf{x}^* \rightarrow \{\underline{e}\}$

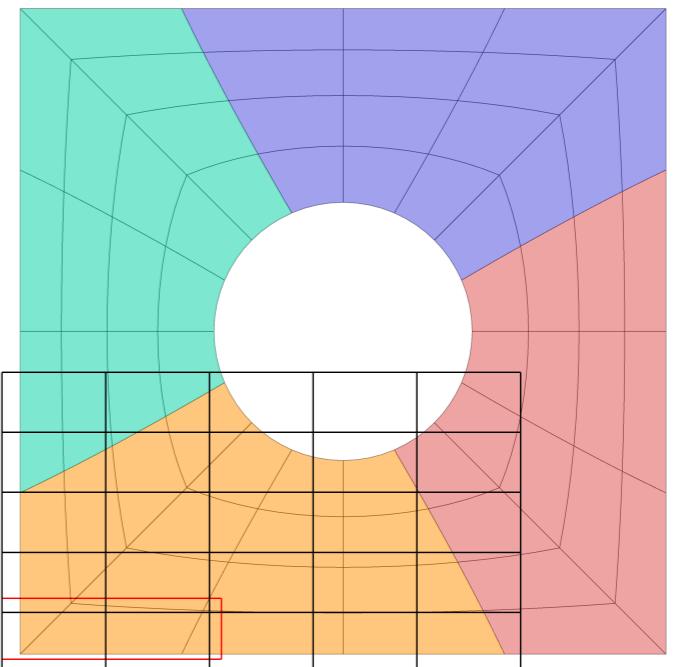
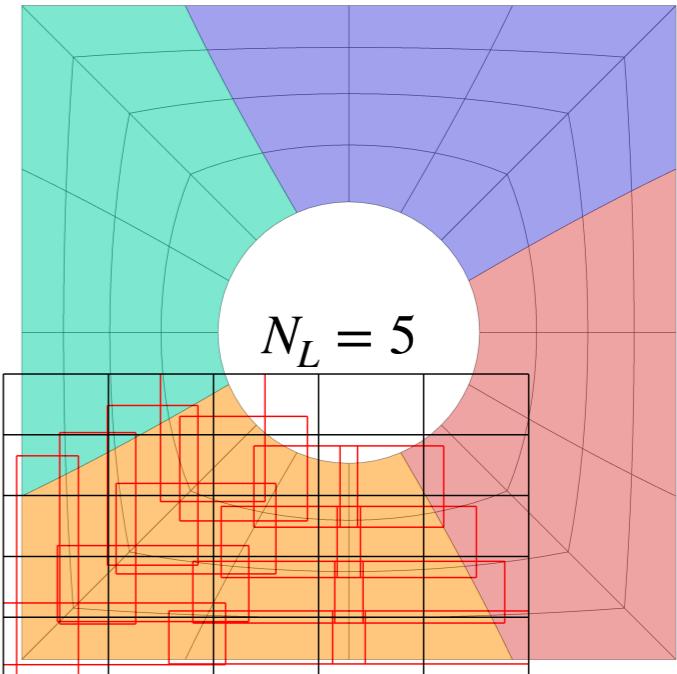
- Generate a uniform $N_L \times N_L$ Cartesian mesh (\mathcal{M}_L) over the domain of the union of processor-local AABB

$\{x_{L,\min}, x_{L,\max}, y_{L,\min}, y_{L,\max}\}.$

- \mathcal{M}_L is never explicitly constructed.
- Compute intersection between **elements of \mathcal{M} and \mathcal{M}_L** :

```
std :: map < int eML, Array < int > eM >
```

- Intersection of AABB and cells of \mathcal{M}_L is trivial.
- Given \mathbf{x}^* , determine which cell of \mathcal{M}_L it is located in, and look-up candidate element using the map.

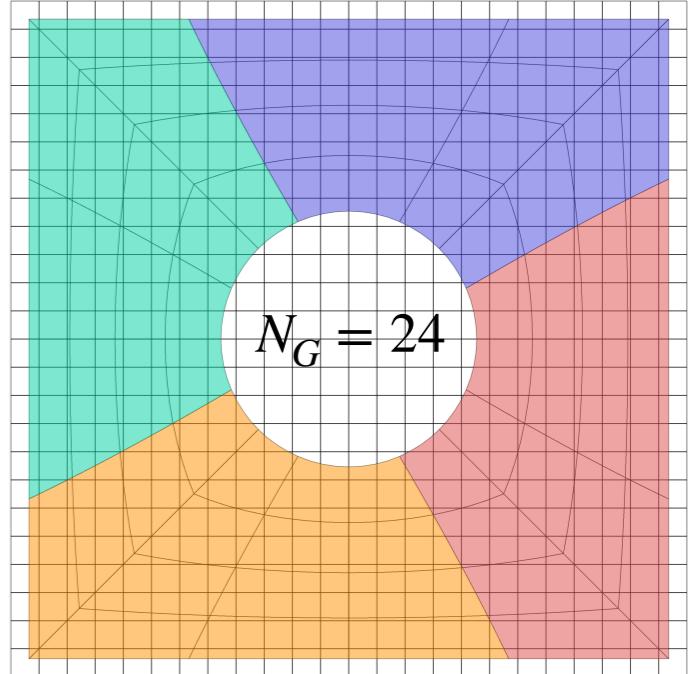


Global Map for $\mathbf{x}^* \rightarrow \{\underline{p}\}$

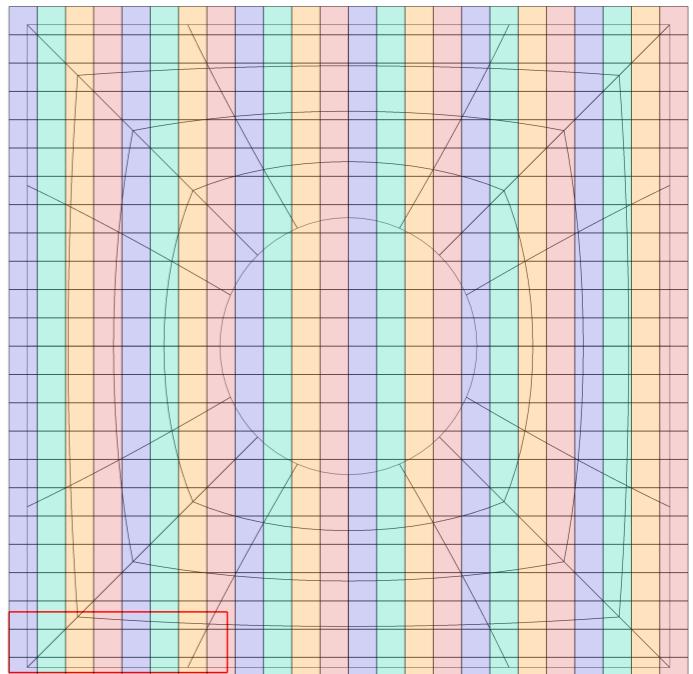
- Generate a uniform $N_G \times N_G$ Cartesian mesh (\mathcal{M}_G) over the entire mesh $\{x_{G,\min}, x_{G,\max}, y_{G,\min}, y_{G,\max}\}$.
- \mathcal{M}_G is globally partitioned. Each rank checks intersection of its elements with elements of \mathcal{M}_G , and communicates to corresponding MPI ranks:

```
std :: map < int eMG, Array < int > pM >
```

- Given \mathbf{x}^* , determine which cell of \mathcal{M}_G it is located in, and send to corresponding MPI rank. Then locally look-up the list of ranks that can contain the point and forward the point to those ranks.



Cartesian mesh overlapping the entire domain.



AABB for one of the elements of \mathcal{M} .



FindPoints



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CPU↔GPU Data Movement

- Data Movement:

[input]

- Mesh nodes - $D \cdot N_E \cdot N^D$

- *Should already be on GPU.*

- Bounding boxes - $(3D + D^2)N_E$

- Local hash mesh - $O(N_E^D)$

- Coordinates of points to be found - $D \cdot N_{pt}$

[output]

- Element index - N_{pt}

- Reference-space coordinates - $D \cdot N_{pt}$

- Distance between actual and found point - N_{pt}

- code (inside element/border/not found) - N_{pt}

- Coalesced memory access.

- Shared memory for fast data access of work arrays used with SIMD instructions.



Initial Element Look-up and Bounding Box Test

- For each point, look up candidate elements based on the hash table. Then check bounding boxes for each candidate elements:
 - AABB test
 - $(x_i^* - x_{i,\min}^e)(x_{i,\max}^e - x_i^*) > 0 \quad \forall i \in [1,D], i \in \mathbb{Z}$
 - OBB test
 - $-1 \leq A_{D \times D}^{-1}(\mathbf{x}^* - \mathbf{x}_c) \leq 1$, where $A_{D \times D}$ captures the OBB size and orientation, and \mathbf{x}_c is the OBB center.

FindPoints - Newton's Method

- Minimize $f(\mathbf{r}) = \frac{1}{2} \|\mathbf{x}^* - \mathbf{x}(\mathbf{r})\|_2^2 = \frac{1}{2} \|\Delta \mathbf{x}_i\|_2^2$ using Newton's method with trust-region.
- \mathbf{r}_0 based on closest mesh node.

$$\mathbf{r}_{l+1} = \mathbf{r}_l - \underbrace{\mathcal{H}_{ji}^{-1} \mathcal{J}_j}_{\Delta \mathbf{r}_l}, \quad \mathbf{r}_{l+1} \in [-1,1]^D, \quad \Delta \mathbf{r}_l \in \alpha_l [-1,1]^D.$$

$$\mathcal{J}_j = \frac{\partial f}{\partial r_j} = \sum_i \Delta x_i \left(-\frac{\partial x_i}{\partial r_j} \right) = -G_{ij}^T \Delta x_i, \quad \mathcal{H}_{jk} = \frac{\partial^2 f}{\partial r_j \partial r_k} = G_{ji} G_{ik} - \beta \Delta x_i \frac{\partial^2 x_i}{\partial r_j \partial r_k}, \quad G_{ij} = \frac{\partial x_i}{\partial r_j}$$

- $\beta = 0$ when searching inside the element, 1 if searching on element edge/face.
- α_l is a trust-region factor that depends on the quality of most recent Newton update.

Newton's Method - Searching interior to an element

- Ignoring the second derivative term in the Hessian simplifies the Newton update to

$$\mathbf{r}_{l+1} = \mathbf{r}_l + G_{ij}^{-1} \Delta x_i.$$

- Requires evaluation of $\Delta x_i(\mathbf{r}_l) = x_i^* - x_i(\mathbf{r}_l)$ and $G_{ij}(\mathbf{r}_l) = \frac{\partial x_i(\mathbf{r}_l)}{\partial r_j}$.
- We use N_{pt} thread blocks (1 for each point to be found), and each block has $N \cdot D$ threads.
 - Tensor-product structure is leveraged to use 1D operator, which maps well to the N threads.

SIMD Instructions for Parallelizing Work

- First evaluate basis functions and their derivatives:

- N threads - $\phi_i(r)$ and $\frac{\partial \phi_i(r)}{\partial r}$.

- N threads - $\phi_i(s)$ and $\frac{\partial \phi_j(s)}{\partial s}$.

- Next, evaluate the inner summation:

- N threads - $x_i = \sum_{j=1}^N x_{ij}^e \phi_i(r) \phi_j(s)$. Same for $\frac{\partial x_i}{\partial r}, \frac{\partial x_i}{\partial s}$.

- N threads - $y_i = \sum_{j=1}^N y_{ij}^e \phi_i(r) \phi_j(s)$. Same for $\frac{\partial y_i}{\partial r}, \frac{\partial y_i}{\partial s}$.

- Finally, thread 0 accumulates the outer summation.

$$x^e(\mathbf{r}) = \sum_{i=1}^N \sum_{j=1}^N x_{ij}^e \phi_i(r) \phi_j(s), \quad y^e(\mathbf{r}) = \sum_{i=1}^N \sum_{j=1}^N y_{ij}^e \phi_i(r) \phi_j(s)$$

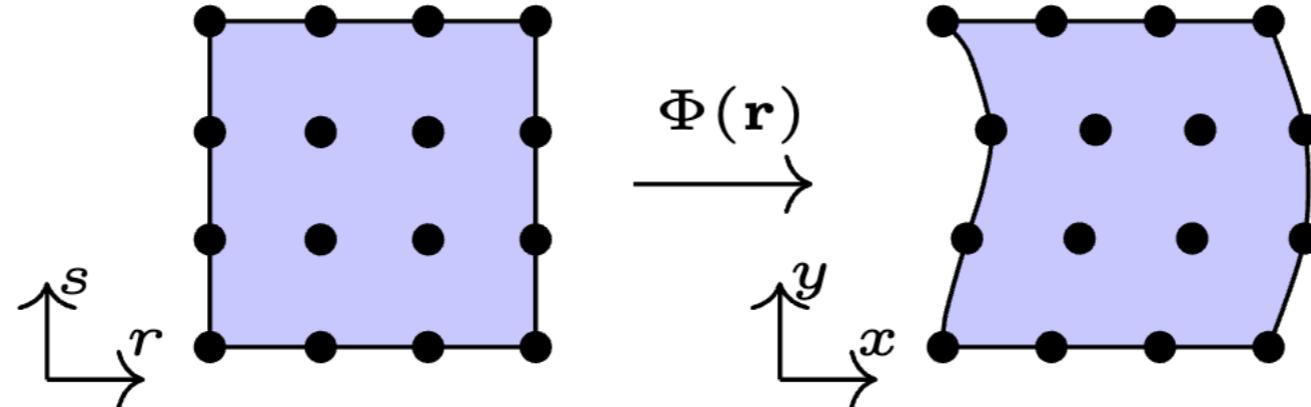
$$\frac{\partial x}{\partial r} = \sum_{i=1}^N \sum_{j=1}^N x_{ij}^e \frac{\partial \phi_i(r)}{\partial r} \phi_j(s), \quad \frac{\partial y}{\partial r} = \sum_{i=1}^N \sum_{j=1}^N y_{ij}^e \frac{\partial \phi_i(r)}{\partial r} \phi_j(s)$$

$$\frac{\partial x}{\partial s} = \sum_{i=1}^N \sum_{j=1}^N x_{ij}^e \phi_i(r) \frac{\partial \phi_j(s)}{\partial s}, \quad \frac{\partial y}{\partial s} = \sum_{i=1}^N \sum_{j=1}^N y_{ij}^e \phi_i(r) \frac{\partial \phi_j(s)}{\partial s}$$



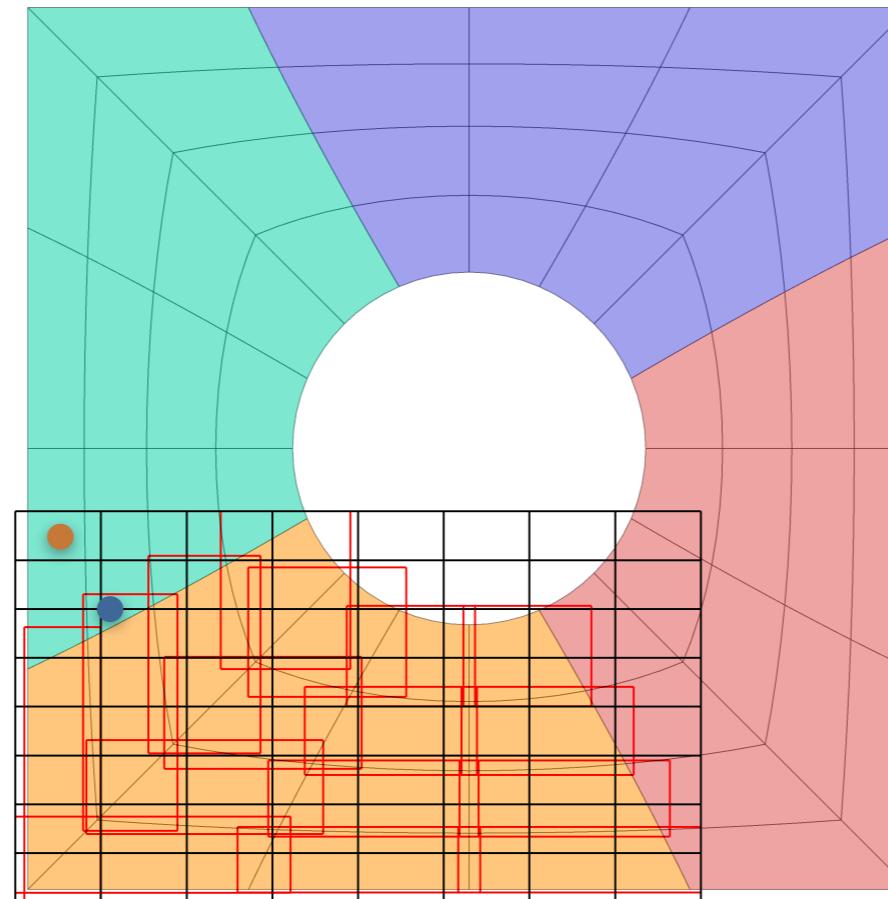
Searching on Element Face/Edge

- After the first Newton iteration, we check \mathbf{r}_l to search either inside an element, on the face (in 3D), or on the edge of the element.
- Fewer unconstrained variables when searching on face (2 in 3D) or edge (1).
- For example, $\mathcal{J} = \frac{\partial f}{\partial r}$ and $\mathcal{H} = \frac{\partial^2 f}{\partial^2 r}$ for edge corresponding to $s \pm 1$.



Logic for Points Not Found Locally

- Points that are not found in the local mesh are routed to other MPI ranks using the global map.
- For points found outside the element, we use element that returns minimum $||\Delta x||$.



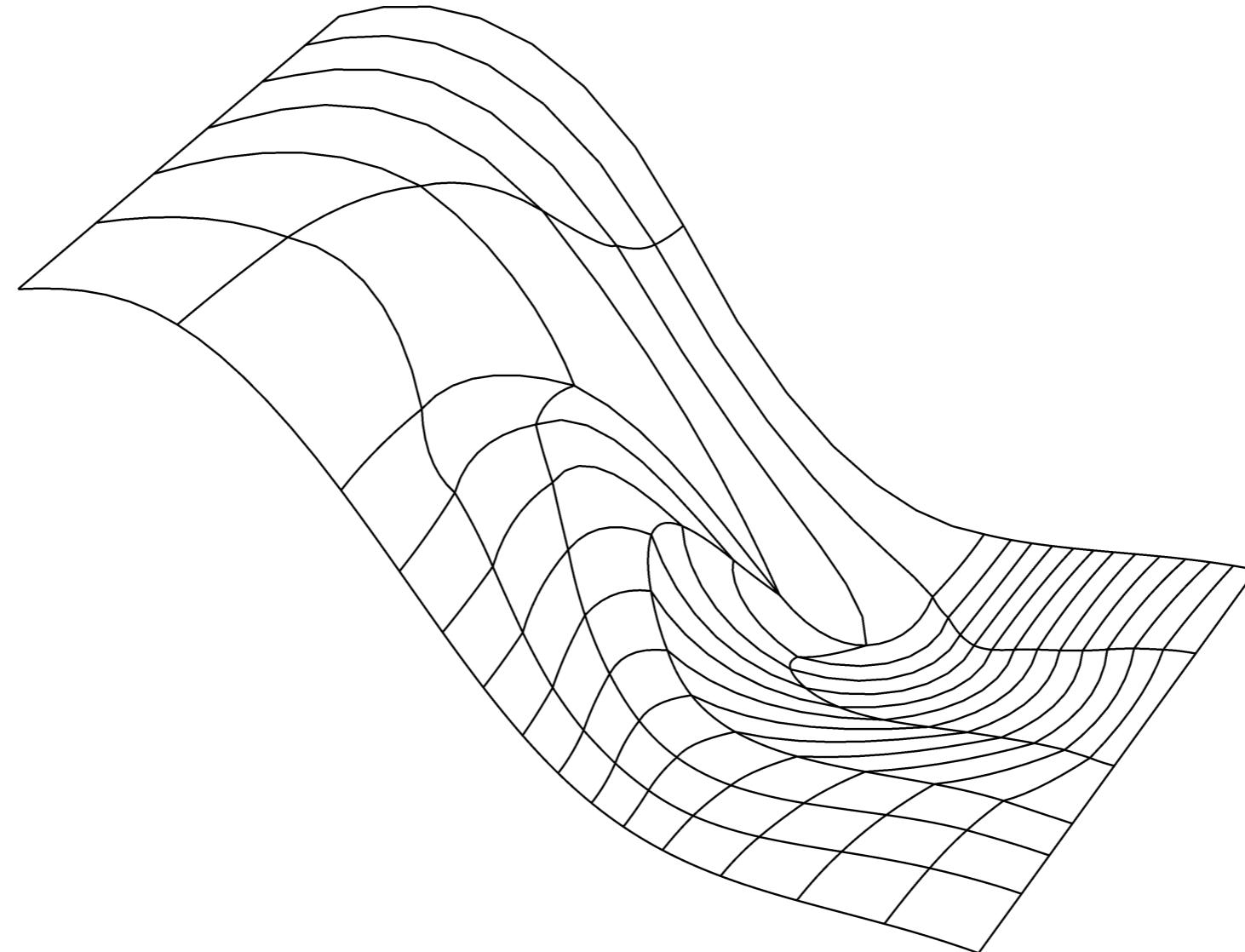
- Point inside the hash mesh but not in any element's bounding box.
- Point inside the bounding box but not the element

Interpolation

$$u(\mathbf{x}(\mathbf{r})) = u(\mathbf{r})|_{\Omega^e} = \sum_{i=1}^{\tilde{N}} \sum_{j=1}^{\tilde{N}} \sum_{k=1}^{\tilde{N}} u_{ijk}^e \phi_i(r) \phi_j(s) \phi_k(t),$$

- N_{pt} blocks, 1 for each point, with \tilde{N}^d threads in each block.
- Interpolation done first for processor local points, and then for points originating on other ranks but owned by elements on current rank.

Extension to Surface Meshes



Aditya Parik's talk at 1:20 PM

Results



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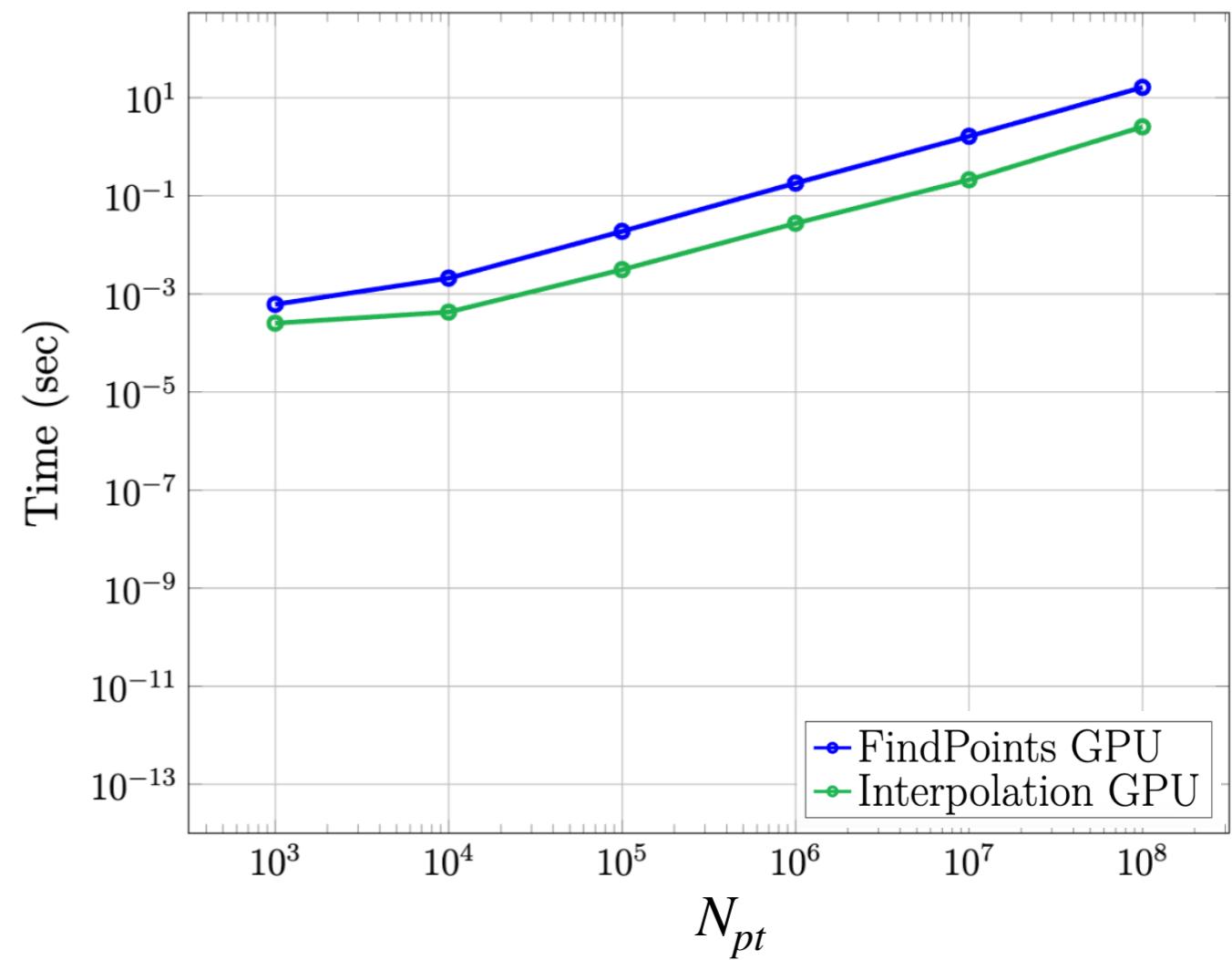
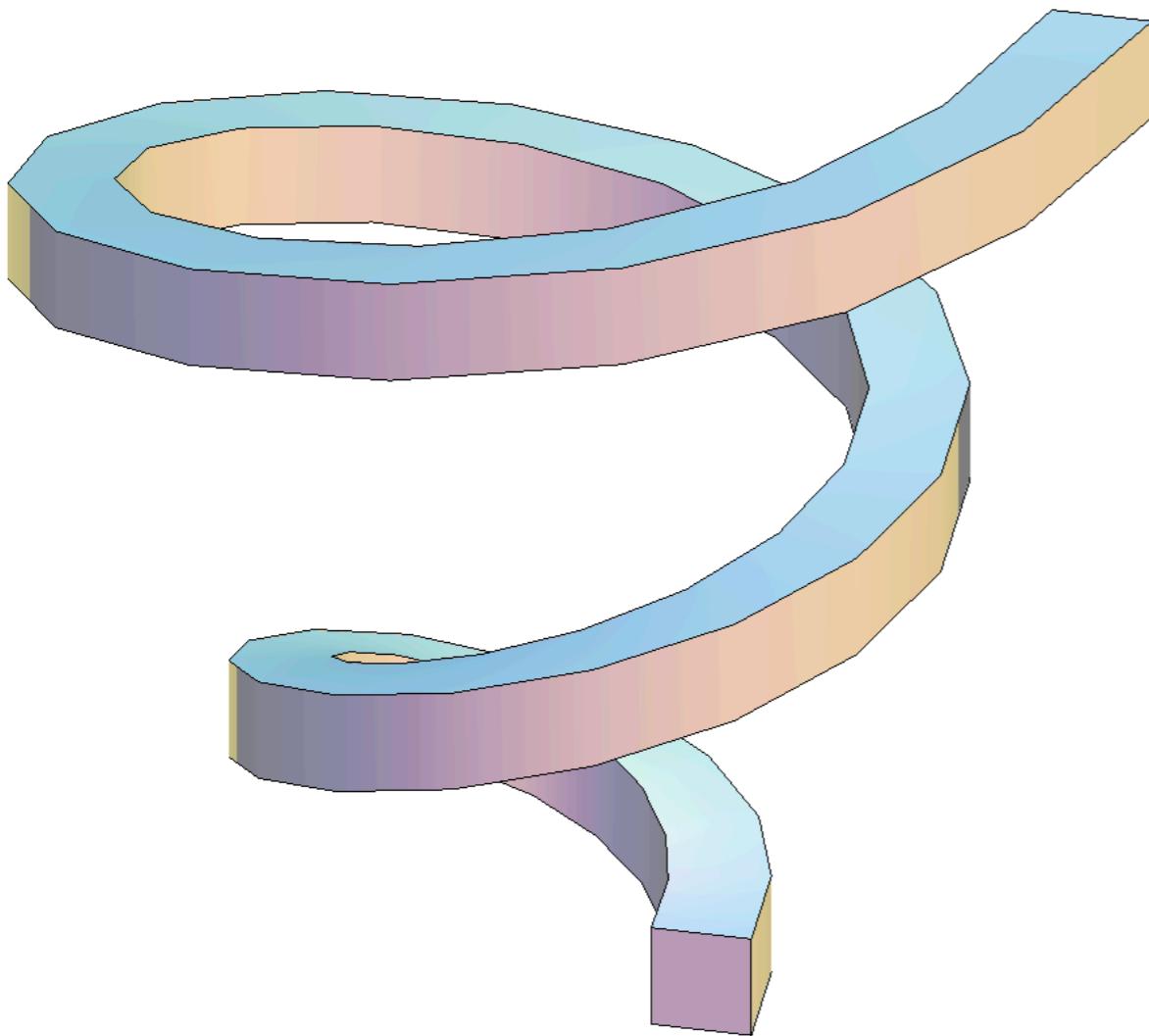
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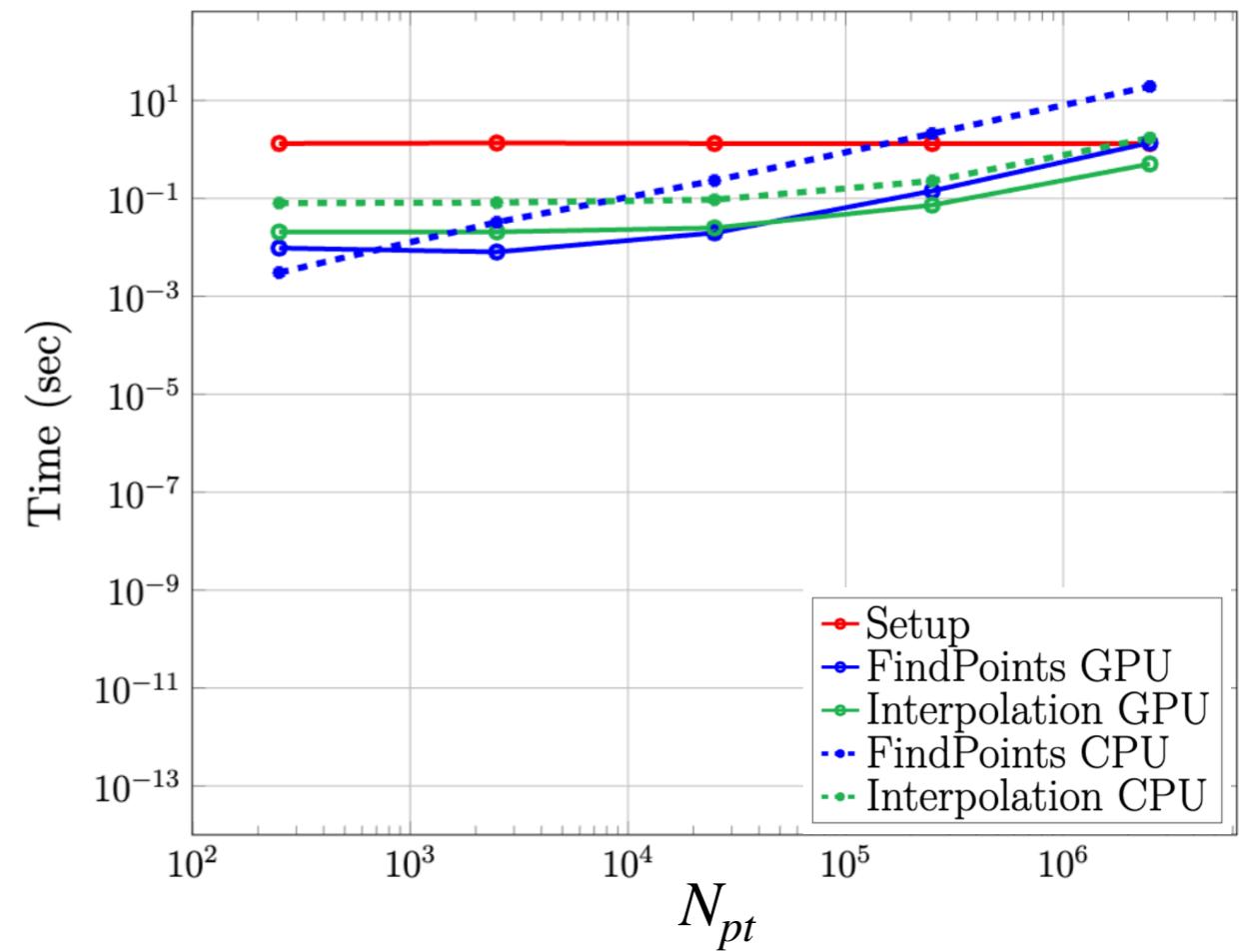
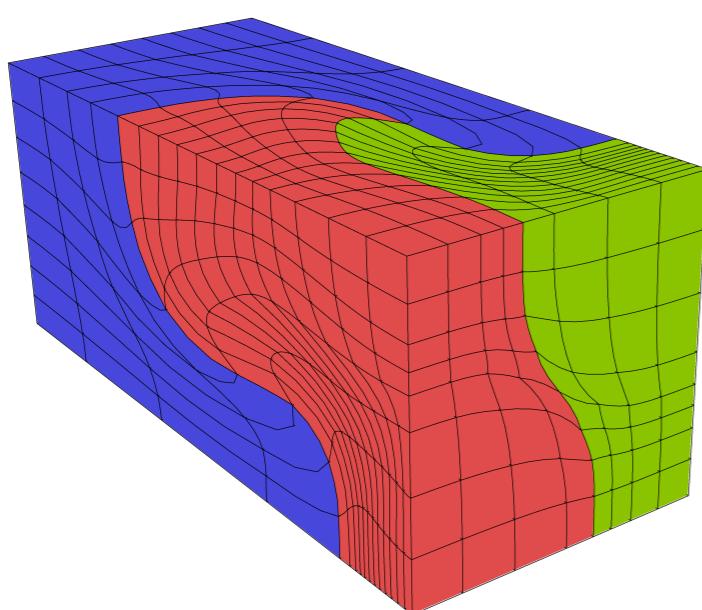
Spiral ($p = 9$)

- Time to find up-to 100 million points in a 9th order element on 1 GPU ([miniapps/meshing/pfindpts](#)).
- 5 Newton iterations per point on average.



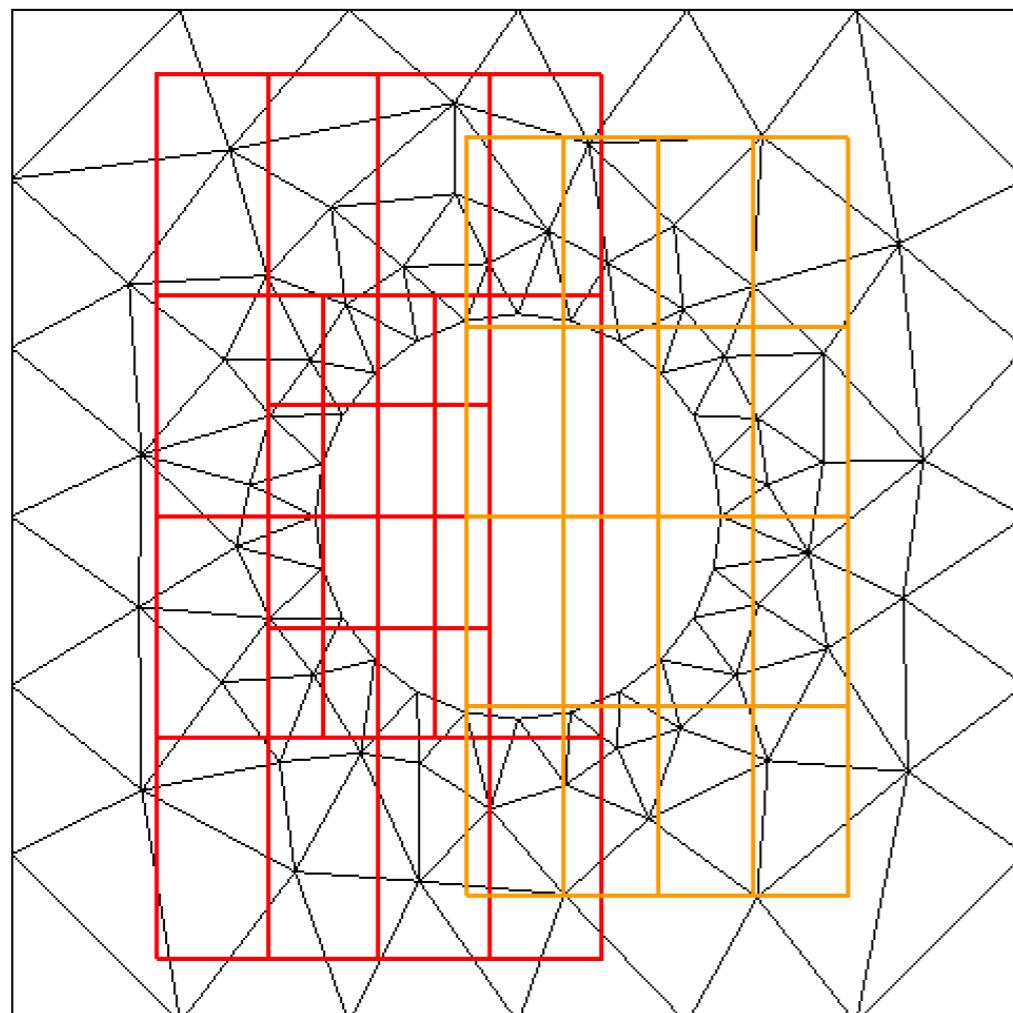
Triple Point Problem ($p = 3$)

- $N_E = 65,536, p = 3$ for the triple point problem on 4 GPUs.
- Lassen supercomputer @ LLNL
 - 756 Nodes: 40 IBM Power9 CPU Cores + 4 Nvidia V100 GPUs per Node.
 - In GPU mode, we typically run on 4 GPUs + 4 CPU cores with all computation on GPUs.

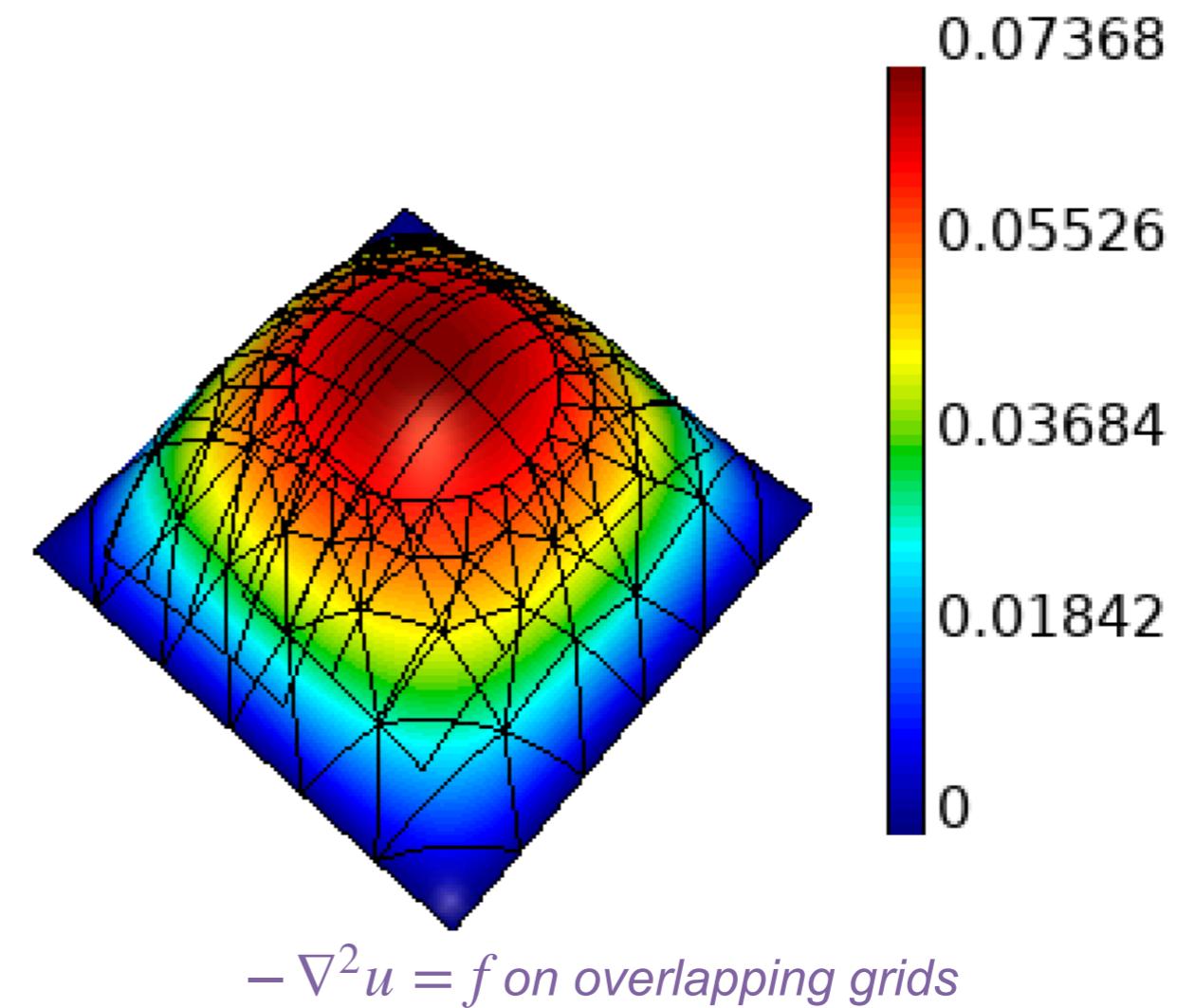


Applications - Solving PDEs on Overlapping Grids

- miniapps/gslib/schwarz_ex1p



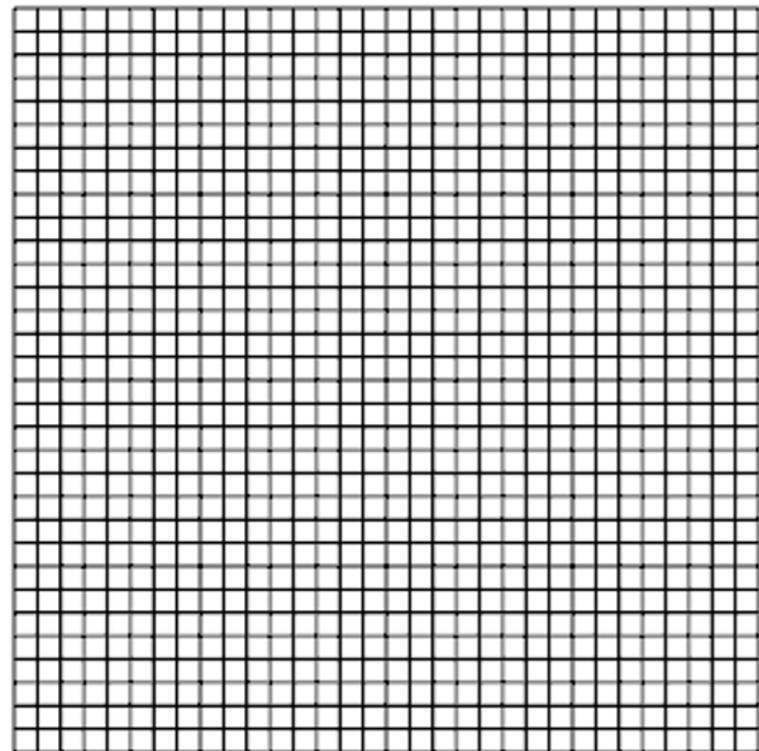
Overlapping grids



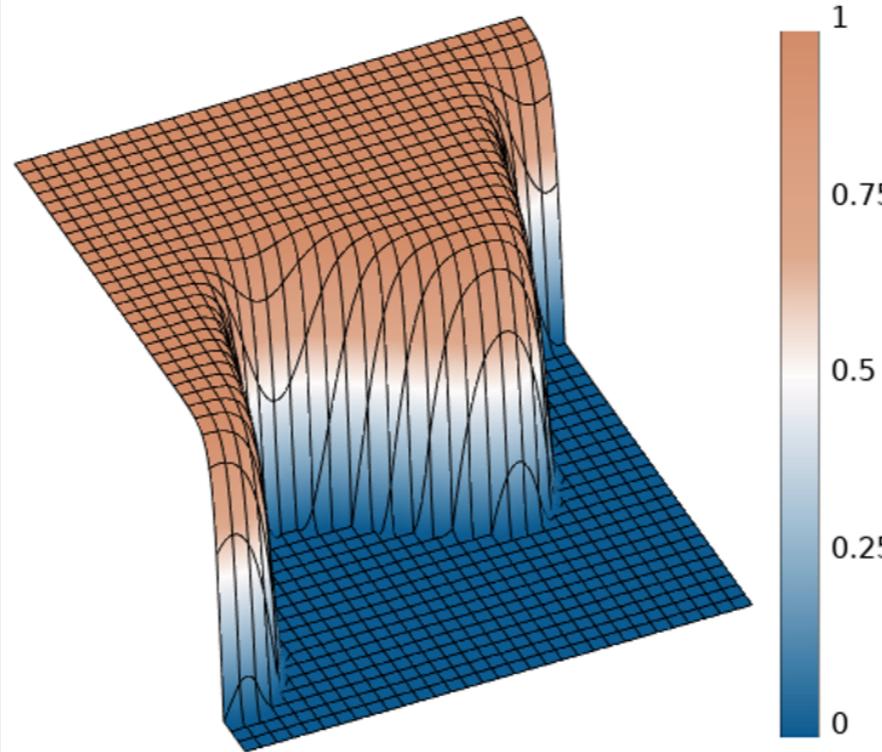
$-\nabla^2 u = f$ on overlapping grids

Applications - Mesh to Mesh Remap during r-adaptivity

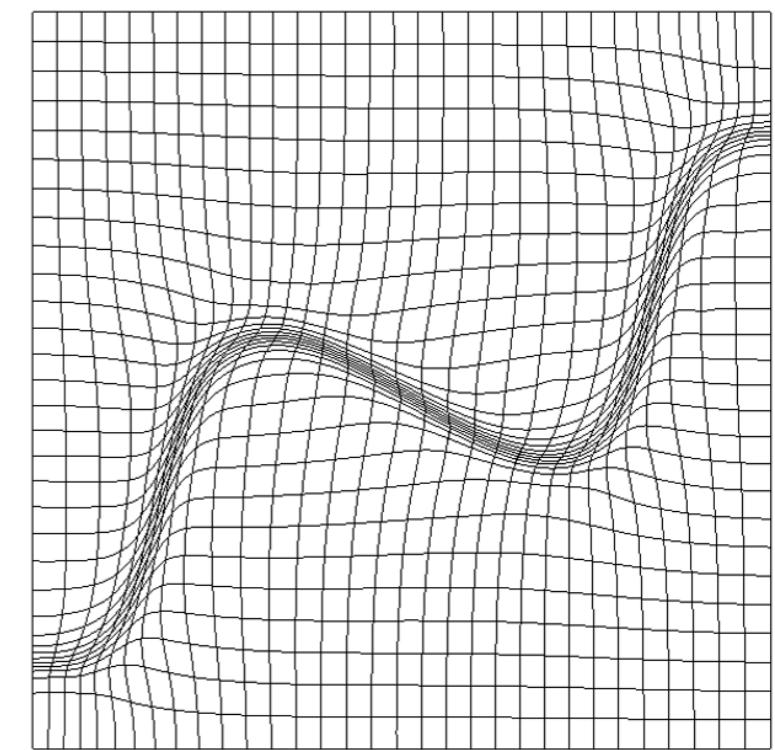
- miniapps/meshing/pmesh – optimizer



Initial mesh



Discrete function on initial mesh



Mesh adapted to discrete function with GSLIB-based remap during mesh optimization.

Some Remarks on Usage!

- All MPI ranks must call the methods simultaneously.
- Do not duplicate list of points on all the ranks.
- New methods enable custom interpolation where we do not directly use a GridFunction.

```
FindPointsGSLIB::finder(MPI_COMM_WORLD);
finder.Setup(pmsh);
if ((myid != 0)) { xyz.Destroy(); }
finder.FindPoints(xyz, point_ordering);
finder.Interpolate(gf_in, interp_xyz);

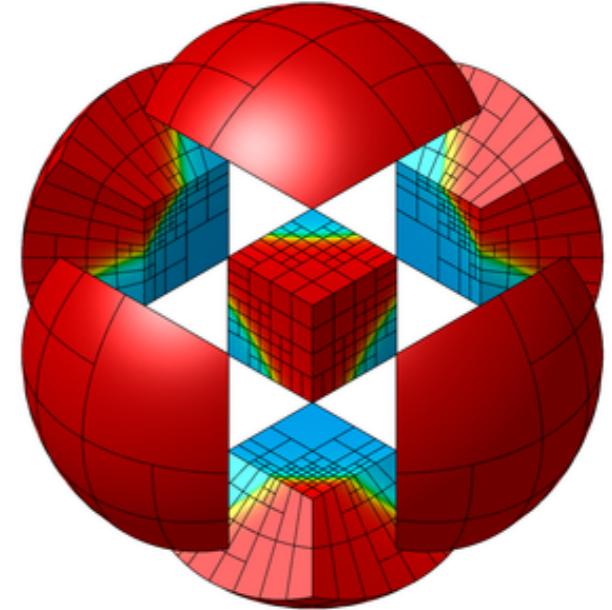
· FindPointsGSLIB::finder;
· finder.Setup(pmsh);
· finder.FindPoints(xyz, Ordering::byVDIM);

· /* Interpolate gradient using custom interpolation procedure. */
· // We first send information to MPI ranks that own the element corresponding
· // to each point.
· Array<unsigned int> recv_elem, recv_code;
· Vector recv_rst;
· finder.DistributePointInfoToOwningMPIRanks(recv_elem, recv_rst, recv_code);
· int npt_recv = recv_elem.Size();
· // Compute gradient locally
· Vector grad(npt_recv*dim);
·
·
·
· // Send the computed gradient back to the ranks that requested it.
· Vector recv_grad;
· finder.DistributeInterpolatedValues(grad, dim, Ordering::byVDIM, recv_grad);
```

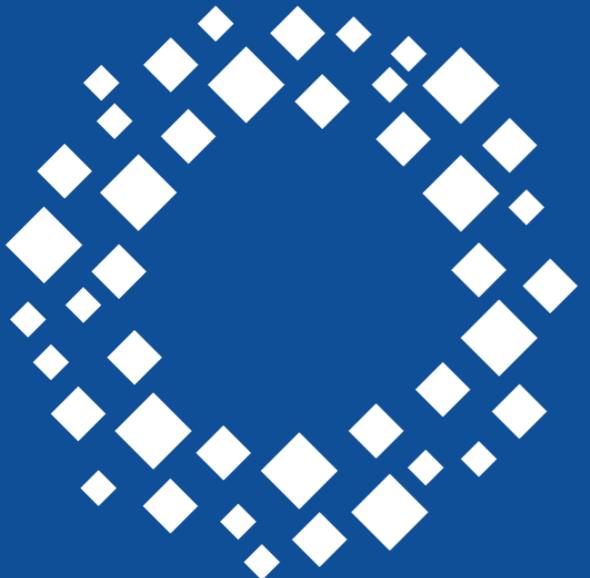


Summary & Future Work

- Thanks to Yohann Dudouit for the discussions.
- Robust arbitrary point search in high-order meshes on GPUs.
- Future work will extend implementation to simplices on GPUs.
- Paper under preparation with all the technical details!



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