Learning topological operations on meshes with application to block decomposition of polygons

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Motivation: High-Order Methods on Unstructured Meshes

- Widely believed that *high-order accurate methods* will be required for challenging simulations (turbulent flows, wave propagation, etc)
- In addition, *fully unstructured meshes* are necessary to handle complex geometries, with adapted resolution and full automation
- Goal: Develop robust, efficient, and accurate high-order methods



A Face Upwinded Spectral Element Method (FUSE) for Conservation Laws

Scientific Achievement

A new stabilization scheme for high-order continuous Spectral Element Methods which is provable convergent up to any order.

Significance and Impact

The work has the potential to drastically improve the performance of high-order methods, which are widely believed to be required for accurate predictions of turbulent flows and problems with waves and non-linear interactions.

Technical Approach

- Most stabilized schemes for fluids and other conservation laws are based on discontinuous formulations (e.g., the discontinuous Galerkin method)
- A remarkably simple way to stabilize continuous methods: Inspired by finite difference methods, choose the full upwind stencil only for face nodes
- Provably high-order convergent for a non-standard node distribution
- In addition, a line-based sparsity patterns bring the Jacobian cost from $\mathcal{O}(p^D)$ to $\mathcal{O}(pD)$, for polynomial degree p in D dimensions

Pl(s)/Facility Lead(s): Per-Olof Persson, LBNL Math Group ASCR Program: Base Math ASCR PM: Steven Lee Publication(s) for this work: Y. Pan, P.-O. Persson, "A Face-Upwinded Spectral Element Method on Unstructured Quadrilateral Meshes," *Journal of Computational Physics* (in review) Y. Pan, P.-O. Persson, "A Stabilized Face-Upwinded High-Order Method for Incompressible Flows," *Proc. of 2023 AIAA AVIATION*, June 2023.





Previous work: HOIST – High-Order Implicit Shock Tracking

- Use full-space optimization to align high-order curved meshes with discontinuities
- Error estimator based on a *p*-adaptive residual:

$$r_{h,p'}^K(U_{h,p}) \coloneqq \int_{\partial K} \psi_{h,p'}^+ \cdot \mathcal{H}(U_{h,p}^+, U_{h,p}^-, n) \, dS - \int_K F(U_{h,p}) : \nabla \psi_{h,p'} \, dV$$

- Minimize $f_{\text{err}}(u, x) \coloneqq \frac{1}{2} \mathbf{R}(u, x)^T \mathbf{R}(u, x)$ to align mesh to shocks
- Solve with efficient SQP solver based on a full-space approach



Zahr, Persson, (JCP 2018), Zahr, Shi, Persson (JCP 2020/22), Zahr et al (multiple).

Deep Reinforcement Learning for Block Meshing

- Define a "game" for automatic block mesh improvement:
 - "Moves": Local or global topological operations (e.g. "flips")
 - "Score": Measure of irregularity of the mesh $s = \sum |\Delta_i|$
- Use a half-edge mesh structure to define a CNN-type network which extends to fully unstructured quadrilateral meshes
- Train on random geometries, using the PPO algorithm on GPUs
- Consistently produces close-to-optimal meshes



[1] Narayanan, Pan, Persson. *Learning topological operations on meshes with application to block decomposition of polygons*. In review & arXiv:2309.06484.

Live Mesh Demo

Basic idea of reinforcement learning



Reinforcement Learning, Solutions Methods









Sampling based methods

- Monte Carlo Tree Search
- Deep RL

 $\Pi(a_t|s_t|\theta)$

Policy: Probability distribution over actions

 $P(s_{t+1}|s_t, a_t)$

State transition probability

$$au = s_0, a_0, \dots, s_H, a_H$$

State - action trajectory

 $R(au) = \sum_{t=0}^{H} R(s_t, a_t)$ Cumulative returns of trajectory

$$U(\theta) = \mathbb{E} \left[R(\tau); \Pi_{\theta} \right]$$
$$= \sum_{\tau} P(\tau; \theta) R(\tau)$$

$$\theta^* = \arg\max_{\theta} U(\theta)$$

Estimating gradient of objective

$$\begin{split} U(\theta) &= \sum_{\tau} P(\tau; \theta) R(\tau) \\ \nabla_{\theta} U(\theta) &= \sum_{\tau} \nabla_{\theta} P(\tau; \theta) R(\tau) \\ &= \sum_{\tau} P(\tau; \theta) \frac{\nabla_{\theta} P(\tau; \theta)}{P(\tau; \theta)} R(\tau) \\ &= \mathbb{E} \left[\nabla_{\theta} \log(P(\tau; \theta)) R(\tau) \right] \quad \approx \frac{1}{m} \sum_{i=1}^{m} \nabla_{\theta} \log(P(\tau^{(i)}; \theta)) R(\tau^{(i)}) \end{split}$$

Mesh editing operations - triangles

Edge-flip

Edge-split

Collapse

Mesh editing operations - quadrilaterals, local



Split-Collapse

Mesh editing operations - quadrilaterals, global





Objective: minimize vertex irregularity

Given:

- Mesh m
- Desired degree of vertices *d**:

 $d^* = \begin{cases} 360/\alpha & \text{interior vertex} \\ \max\left(\lfloor \theta/\alpha \rfloor + 1, 2\right) & \text{boundary vertex} \end{cases}$

where $\alpha = 60$ for triangles, 90 for quads, and θ is the angle of a boundary point.

• Define
$$\Delta_i = d_i - d_i^*$$

minimize $s = \sum_{i} |\Delta_i|$



Note that:

•
$$s^* = \left| \sum_i \Delta_i \right| \le \sum_i |\Delta_i| = s$$

• *s*^{*} is invariant under mesh edits.

This means s^* is a bound on the best possible improved mesh \implies use for a normalized optimality score.



The problem poses several challenges:

- Discrete decisions
- Fully unstructured
- Dynamic data-structure

Solution methods need to be able to:

- Represent and understand mesh topology
- Efficiently implement mesh edits



Half-edges represent topology in a structured way



Half-edge operations used to represent state

Template: Ordered sequence of vertices around each half-edge



In the language of reinforcement learning

- State: Irregularity and degree of vertices in template
- Action: Flip, split, collapse, etc.
- **Reward**: $r_t = s_t s_{t+1}$

Training procedure:

- Generate random 10-30 sided polygons
- Initial mesh by Delaunay refinement, split using Catmull-Clark for quads
- Terminate if $s^* = s$ or a maximum number of steps taken
- Monitor normalized returns

Neural network learns a mesh edit policy



Trained in self-play by Proximal Policy Optimization (PPO) algorithm

Schulman, John, et al. Proximal policy optimization algorithms arXiv:1707.06347 (2017).





Average performance over training history

Evaluating the trained agent on multiple rollouts

Performance of the triangle mesh agent over the training history.

Triangular meshing Example 1 Step 0 (out of 27)



Triangular meshing Example 1 Step 1 (out of 27)



Triangular meshing Example 1 Step 2 (out of 27)



Triangular meshing Example 1 Step 3 (out of 27)



Triangular meshing Example 1 Step 4 (out of 27)



Triangular meshing Example 1 Step 5 (out of 27)



Triangular meshing Example 1 Step 6 (out of 27)



Triangular meshing Example 1 Step 7 (out of 27)



Triangular meshing Example 1 Step 8 (out of 27)



Triangular meshing Example 1 Step 9 (out of 27)



Triangular meshing Example 1 Step 10 (out of 27)



Triangular meshing Example 1 Step 11 (out of 27)



Triangular meshing Example 1 Step 12 (out of 27)



Triangular meshing Example 1 Step 13 (out of 27)


Triangular meshing Example 1 Step 14 (out of 27)



Triangular meshing Example 1 Step 15 (out of 27)



Triangular meshing Example 1 Step 16 (out of 27)



Triangular meshing Example 1 Step 17 (out of 27)



Triangular meshing Example 1 Step 18 (out of 27)



Triangular meshing Example 1 Step 19 (out of 27)



Triangular meshing Example 1 Step 20 (out of 27)



Triangular meshing Example 1 Step 21 (out of 27)



Triangular meshing Example 1 Step 22 (out of 27)



Triangular meshing Example 1 Step 23 (out of 27)



Triangular meshing Example 1 Step 24 (out of 27)



Triangular meshing Example 1 Step 25 (out of 27)



Triangular meshing Example 1 Step 26 (out of 27)



Triangular meshing Example 1 Step 27 (out of 27)



Triangular meshing example: 20-sided polygon



Results: Quadrilateral Meshes





Average performance over training history

Evaluating the trained agent on multiple rollouts

Performance of the quadrilateral mesh agent over the training history.

Block mesh decomposition Example 1 Step 0 (out of 19)



Block mesh decomposition Example 1 Step 1 (out of 19)



Block mesh decomposition Example 1 Step 2 (out of 19)



Block mesh decomposition Example 1 Step 3 (out of 19)



Block mesh decomposition Example 1 Step 4 (out of 19)



Block mesh decomposition Example 1 Step 5 (out of 19)



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Block mesh decomposition Example 1 Step 15 (out of 19)



Block mesh decomposition Example 1 Step 16 (out of 19)



Block mesh decomposition Example 1 Step 17 (out of 19)



Block mesh decomposition Example 1 Step 18 (out of 19)



Block mesh decomposition Example 1 Step 19 (out of 19)


15/1

Block mesh decomposition Example 2 Step 0 (out of 12)



15/1

Block mesh decomposition Example 2 Step 1 (out of 12)



13/1

Block mesh decomposition Example 2 Step 2 (out of 12)



13/1

Block mesh decomposition Example 2 Step 3 (out of 12)



11/1

Block mesh decomposition Example 2 Step 4 (out of 12)



11/1

Block mesh decomposition Example 2 Step 5 (out of 12)



9/1

Block mesh decomposition Example 2 Step 6 (out of 12)



7/1

Block mesh decomposition Example 2 Step 7 (out of 12)



5/1

Block mesh decomposition Example 2 Step 8 (out of 12)



5/1

Block mesh decomposition Example 2 Step 9 (out of 12)



3/1

Block mesh decomposition Example 2 Step 10 (out of 12)



3/1

Block mesh decomposition Example 2 Step 11 (out of 12)

1/1

Block mesh decomposition Example 2 Step 12 (out of 12)



Block decomposition example: 10-sided polygon



Block decomposition example: 20-sided polygon



Block decomposition example: L-shaped domain







Block decomposition example: Star-shaped domain



Block decomposition example: Notch domain







Block decomposition example: Double notch domain







Block decomposition example: Square hole in circle domain



Conclusions

- Representation of mesh topology for neural networks
- Unified method to optimize connectivity of triangular and quadrilateral meshes
- Heuristics-free method that learns rich behavior from self-play
- Future work: Combine with Monte Carlo Tree Search, more complex geometries, different formulations, optimize for element quality, 3D

[1] Narayanan, Pan, Persson. *Learning topological operations on meshes with application to block decomposition of polygons*. In review & arXiv:2309.06484.