## **Motivation**

- Adaptive meshing is crucial for **simulation** 
  - Yet, it often requires human expertise
- A mesh is characterized by its local element density, i.e., its sizing field
- A sizing field thus reconstructs the mesh
- We combine
  - Message Passing Networks (MPNs) and
- automatic online label acquisition

for **iterative mesh generation**, where each mesh guides the next one.

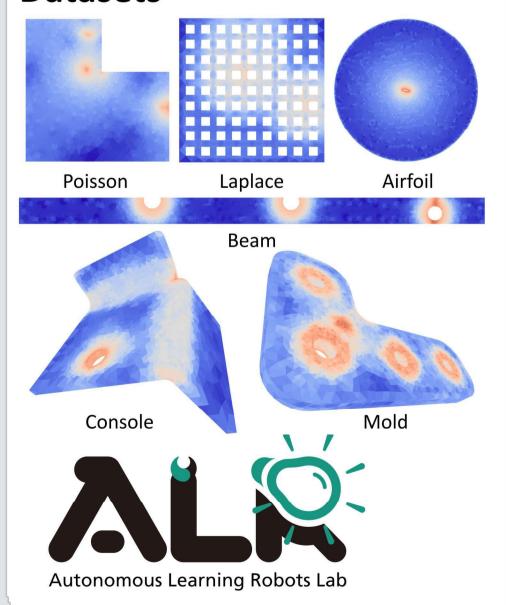
# Inference

- 1. Input an unseen geometry (e.g., an .stl file)
- 2. Generate a coarse uniform mesh
- 3. Iteratively **Encode** the mesh as a graph
  - Predict a sizing field with a MPN
  - Generate an adapted mesh

# **Training**

- 1. Collect a few (~20) expert meshes
- 2. **Generate** coarse uniform meshes
- 3. **Iteratively** o **Project** expert sizing field to mesh
  - Encode the mesh as a graph
  - Train MPN to predict sizing field
  - Generate an adapted mesh
  - Add this mesh to a replay buffer

# **Datasets**

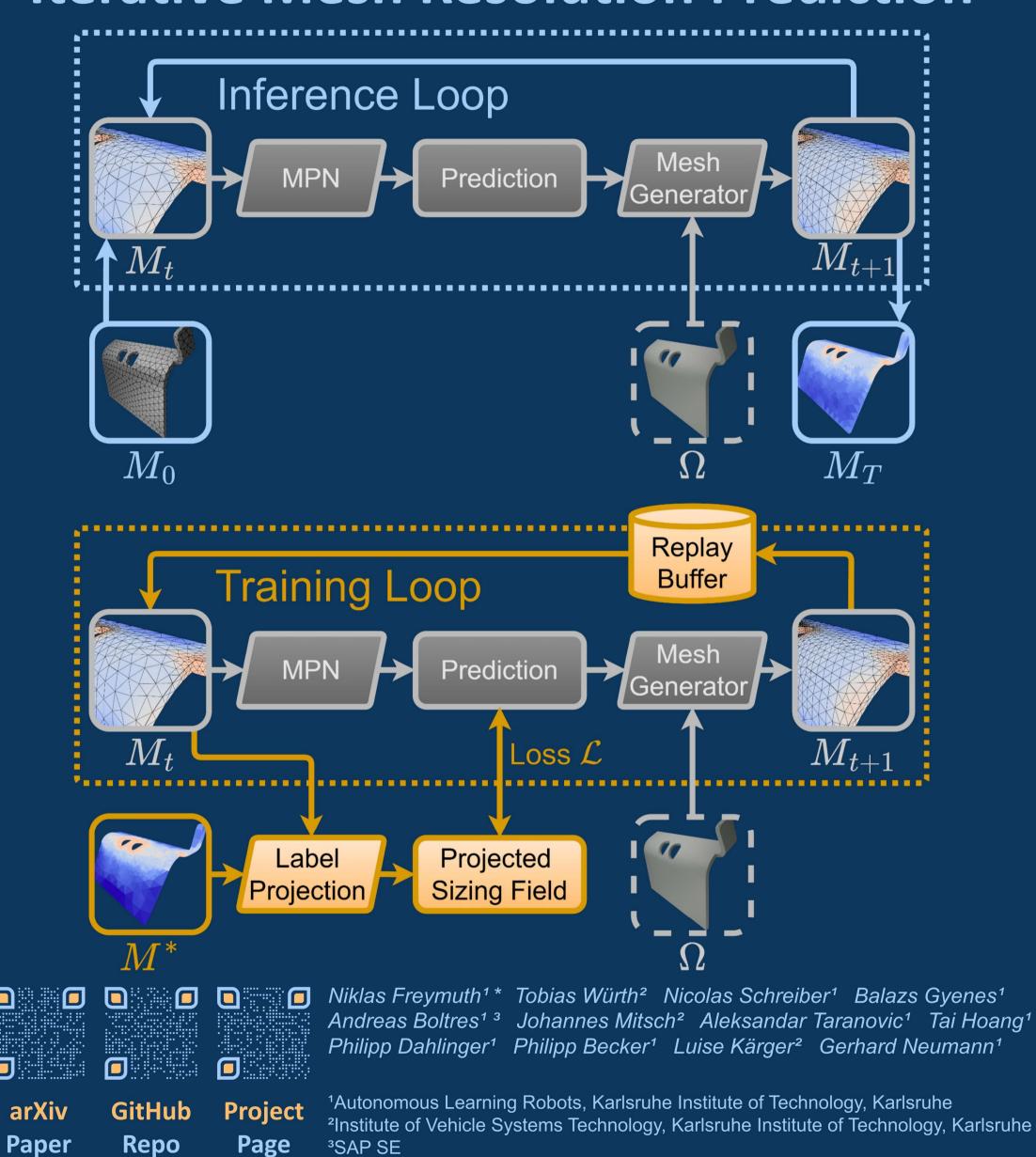


Repo

<sup>3</sup>SAP SE

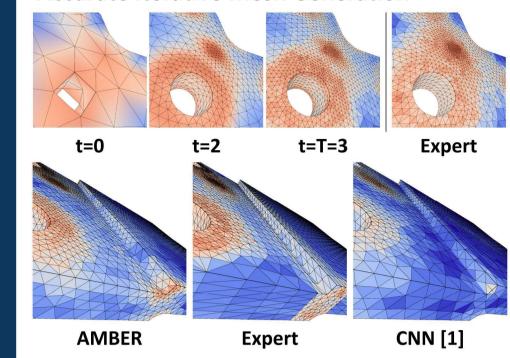
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# **AMBER:** Adaptive Mesh Generation by **Iterative Mesh Resolution Prediction**

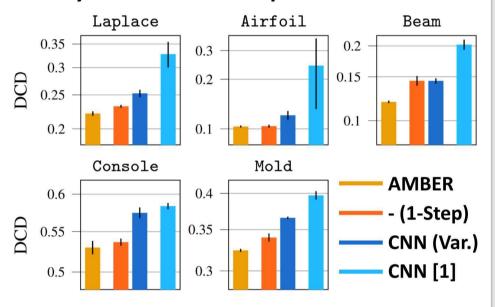


# **Experiments**

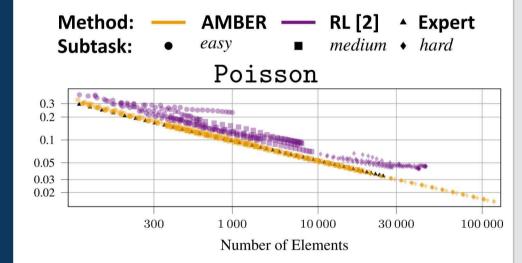
#### **Accurate Iterative Mesh Generation**



### **Closely Match Human Experts**



#### **Zero-Shot Generalization**



#### References

[1] Huang, K., et al. Machine learning-based optimal mesh generation in computational fluid dynamics, 2021. [2] Freymuth, N., et al. Swarm reinforcement learning for adaptive mesh refinement, 2023.

