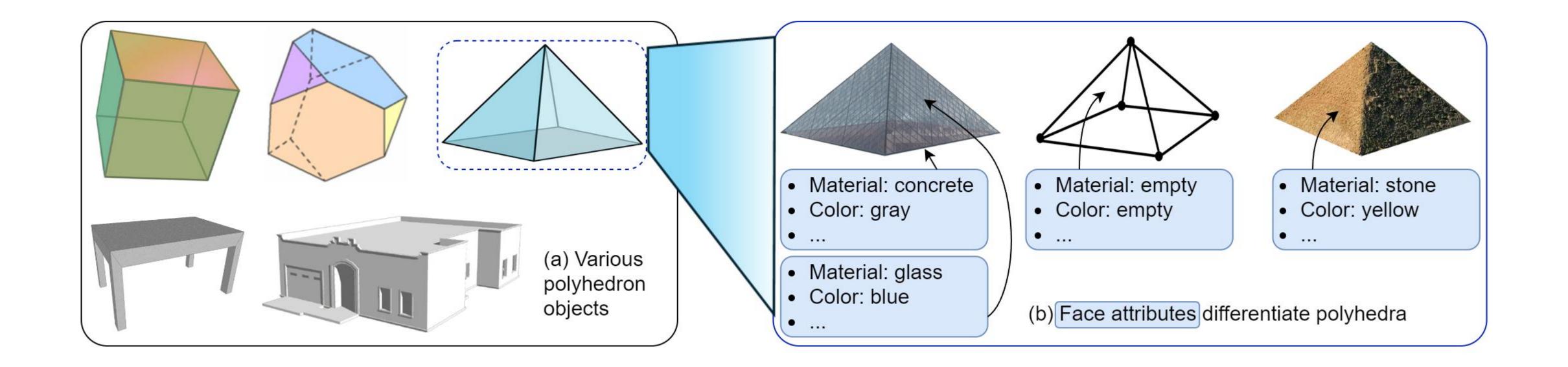


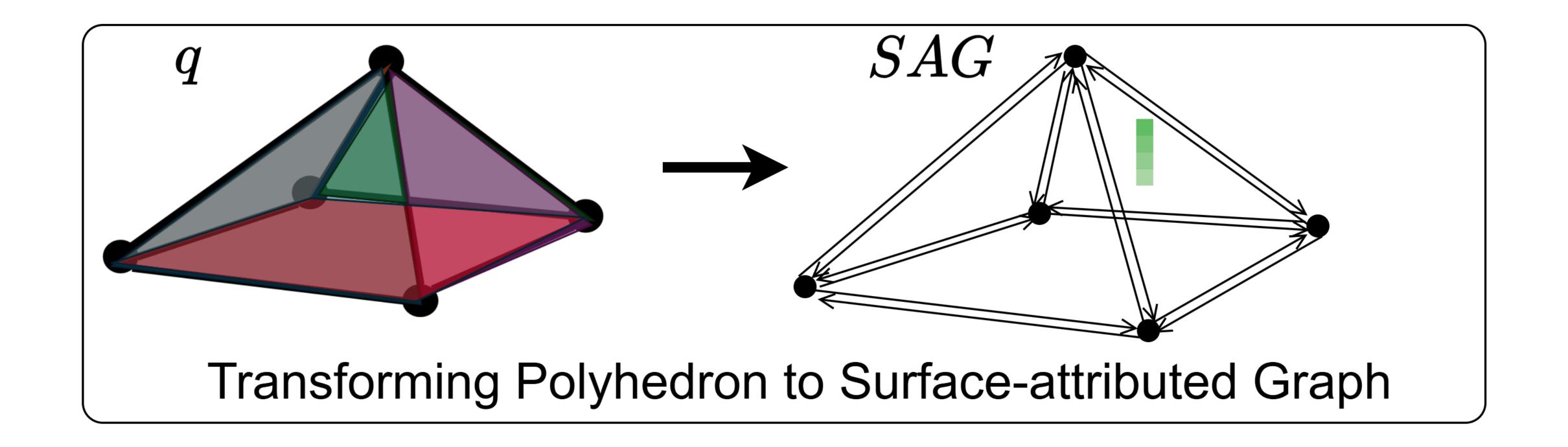
# POLYHEDRONNET: Representation Learning for Polyhedra with Surface-Attributed Graph

Dazhou Yu 2025

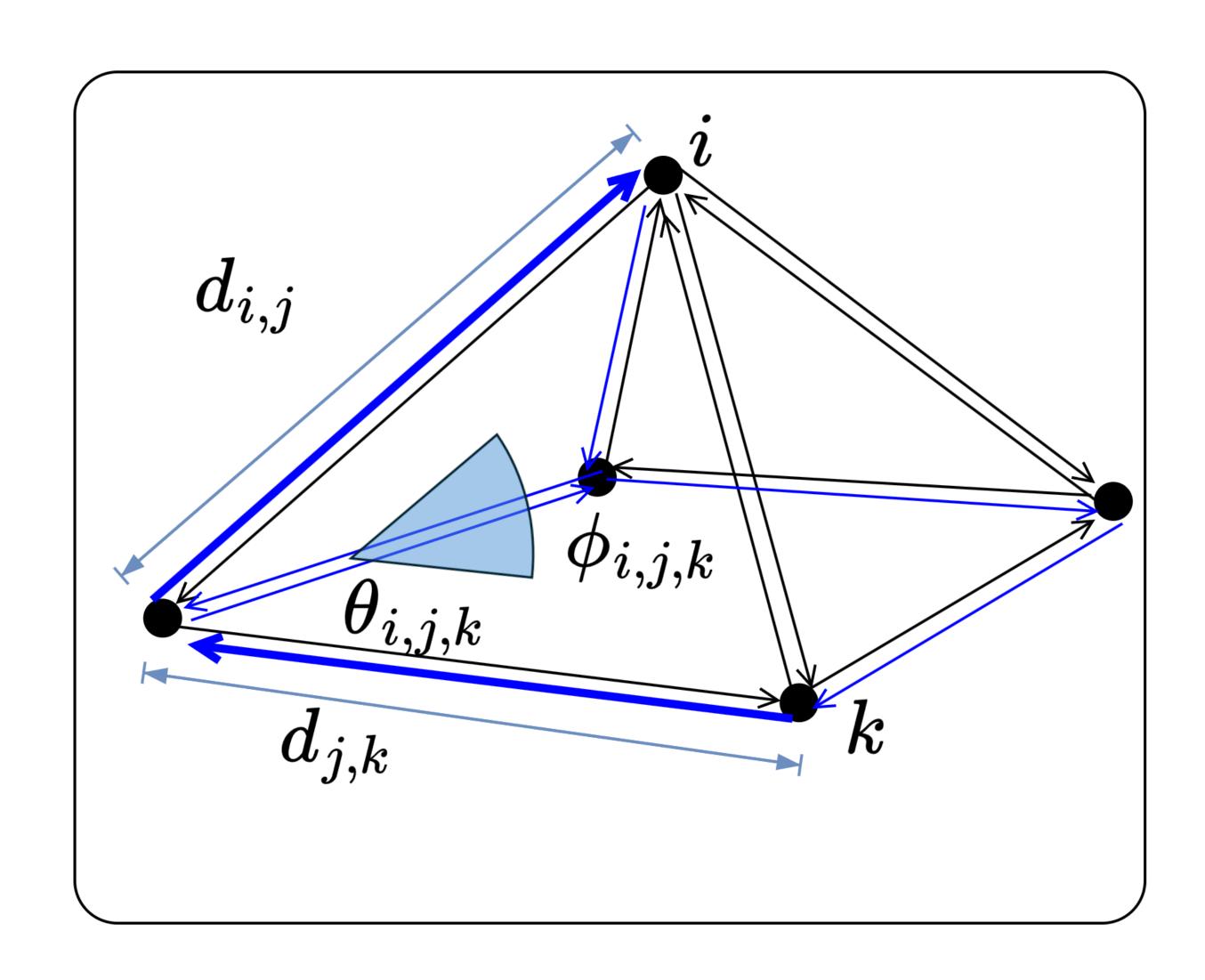
### Polyhedral Representation



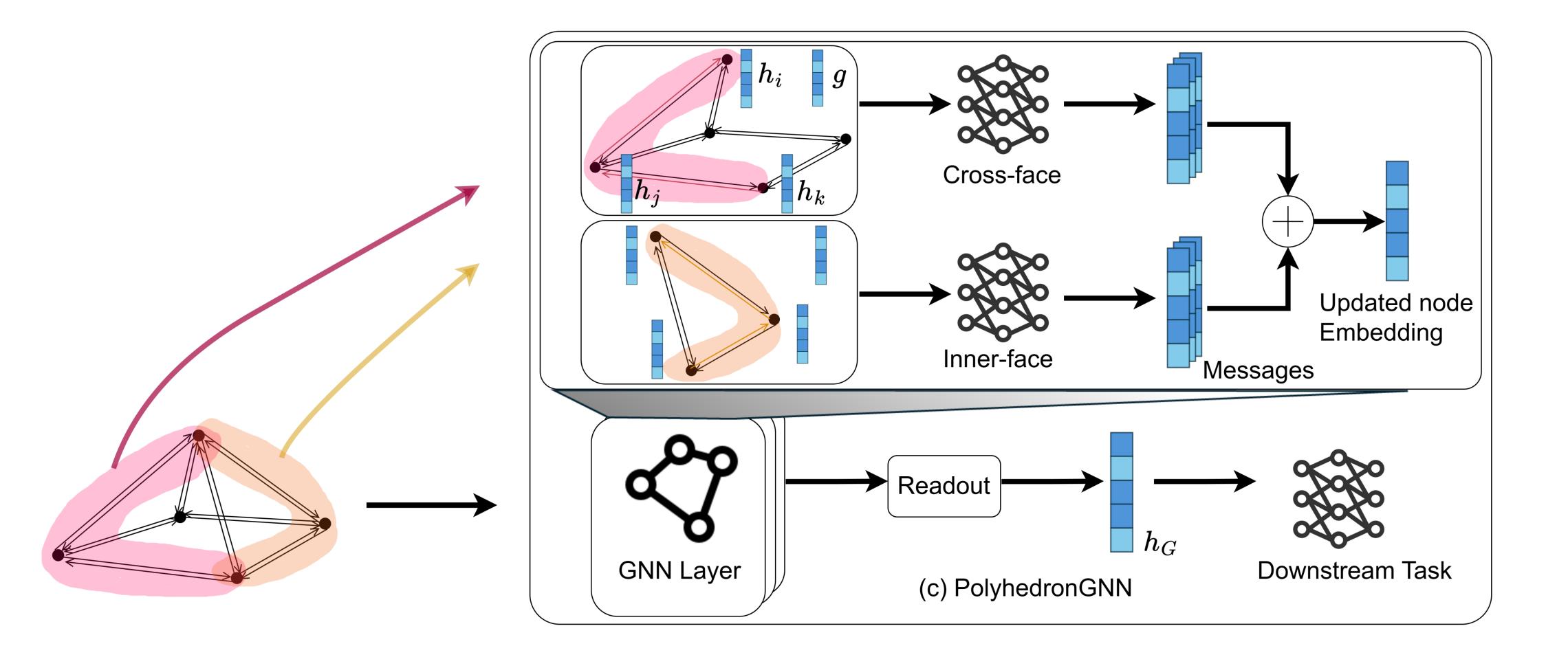
#### Surface-Attributed Graph



# Local Rigid Representation

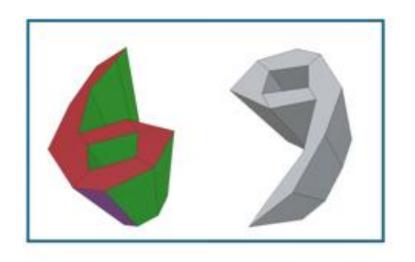


## PolyhedronGNN



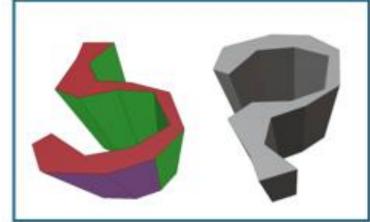
#### Results & Performance

Dataset	Metric	NUFT-DDSL	ResNet1D	NUFT-IFFT	VeerCNN	PolygonGNN	PolyhedronNet
MNIST-C	Acc ↑	0.148	0.152	0.239	0.127	0.435	0.858
	Prec <sup>↑</sup>	0.092	0.139	0.220	0.104	0.446	0.861
	F1↑	0.102	0.083	0.202	0.084	0.427	0.856
	AUC↑	0.474	0.610	0.619	0.576	0.801	0.985
ShapeNet-P	Acc↑	0.097	0.179	0.097	0.163	0.573	0.627
	Prec↑	0.103	0.142	0.082	0.158	0.589	0.640
	F1↑	0.092	0.147	0.083	0.148	0.570	0.625
	AUC↑	0.555	0.625	0.564	0.639	0.916	0.936



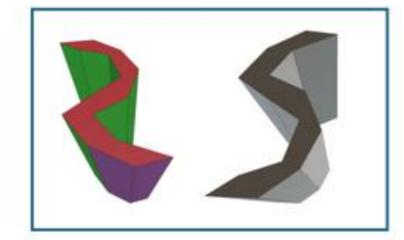
Label: 6

Prediction: 9 6 6 6



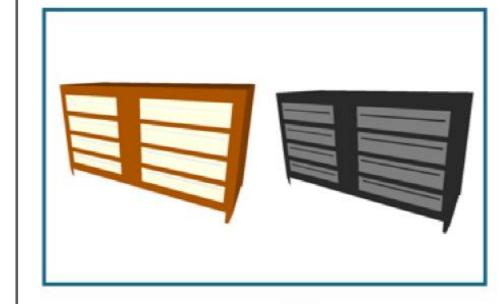
Label: 5

Prediction: 2 5 5 5



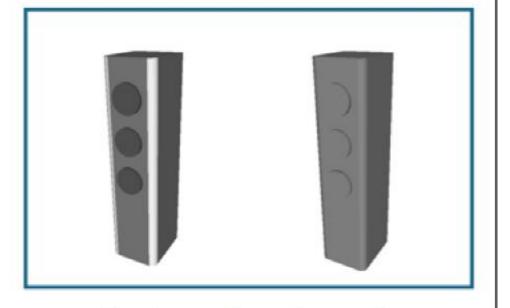
Label: 2

Prediction: 5 2 2 2



Prediction: cabinet

loudspeaker loudspeaker loudspeaker



Prediction: loudspeaker

bookshelf bathtub bookshelf pot

#### Summary & Future Directions

- Powerful unified representation using Surface-Attributed Graph
- Rotation translation invariant using Local Rigid Encoding
- To do:
  - Extend to segmentation
  - Systematic exploration of strengths and weaknesses across various 3D representations polyhedra, point clouds, voxels, multi-view images...



paper



code