

# PolygonGNN: Representation Learning for Polygonal Geometries with Heterogeneous Visibility Graph

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# Background: polygonal representation

- Capturing and encoding the characteristics of **polygonal geometries**
- Polygon  $p$ : an ordered set of coordinates
- A representation learning model on multipolygon is desired

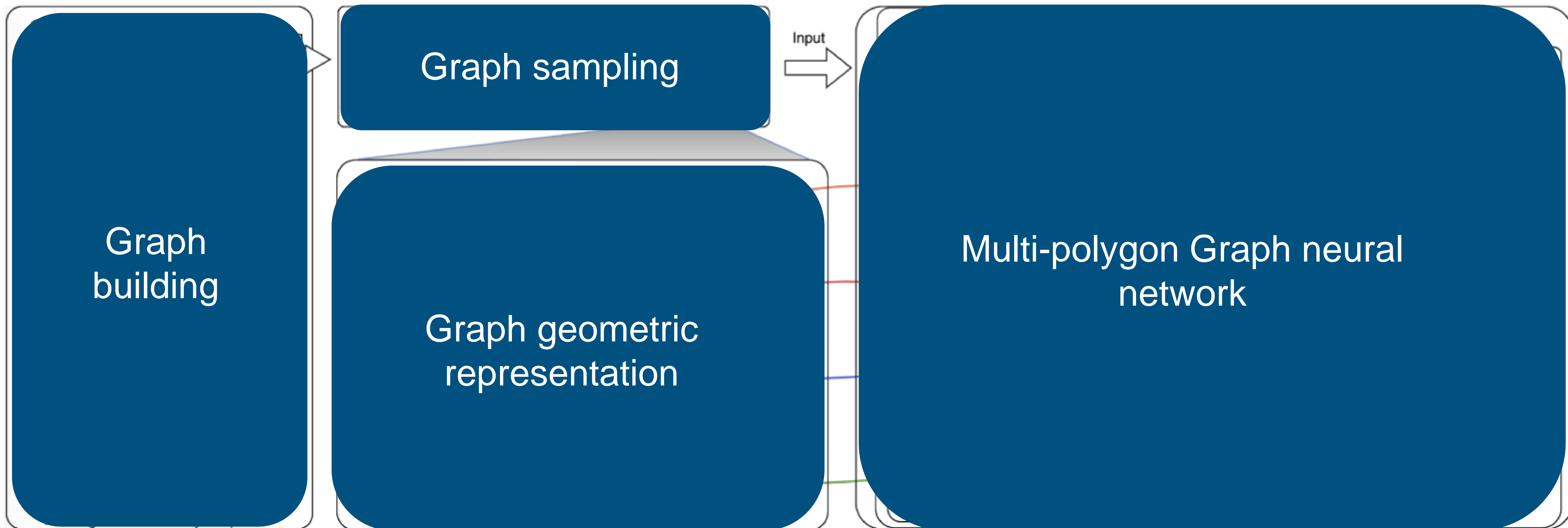


Building Pattern Classification

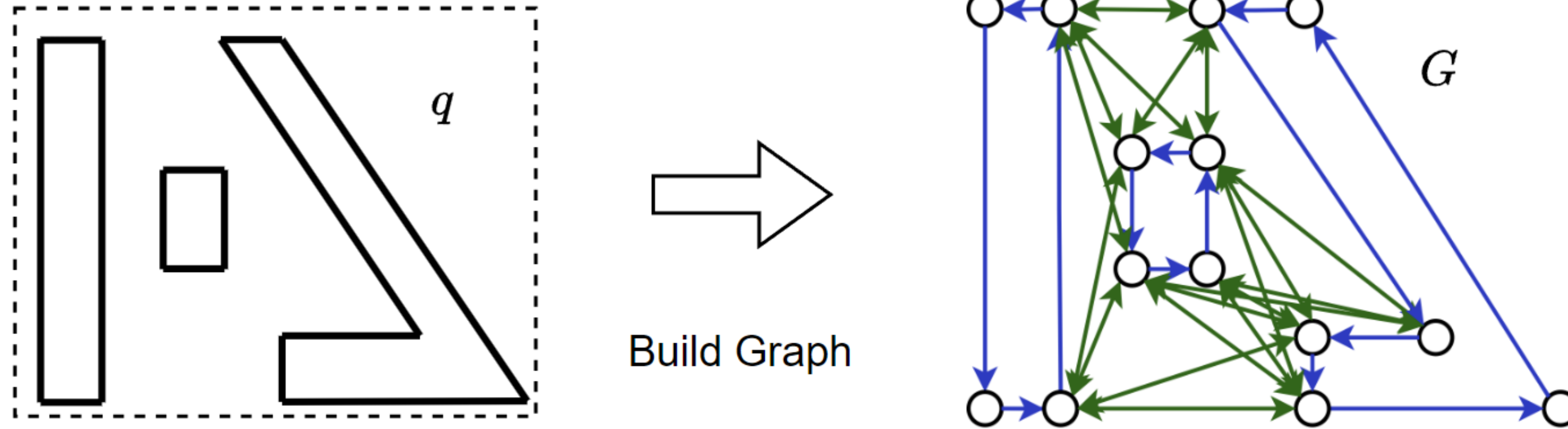
# Challenges & Contributions

- Conserving inner-polygonal & inter-polygonal relationships
  - heterogeneous visibility graphs
- Quadratic complexity from pairwise relationships
  - heterogeneous spanning tree sampling
- Rotation and translation invariance
  - lossless rotation-translation-invariant geometric representation.

# Overview of PolygonGNN Framework



# Heterogeneous Visibility Graphs

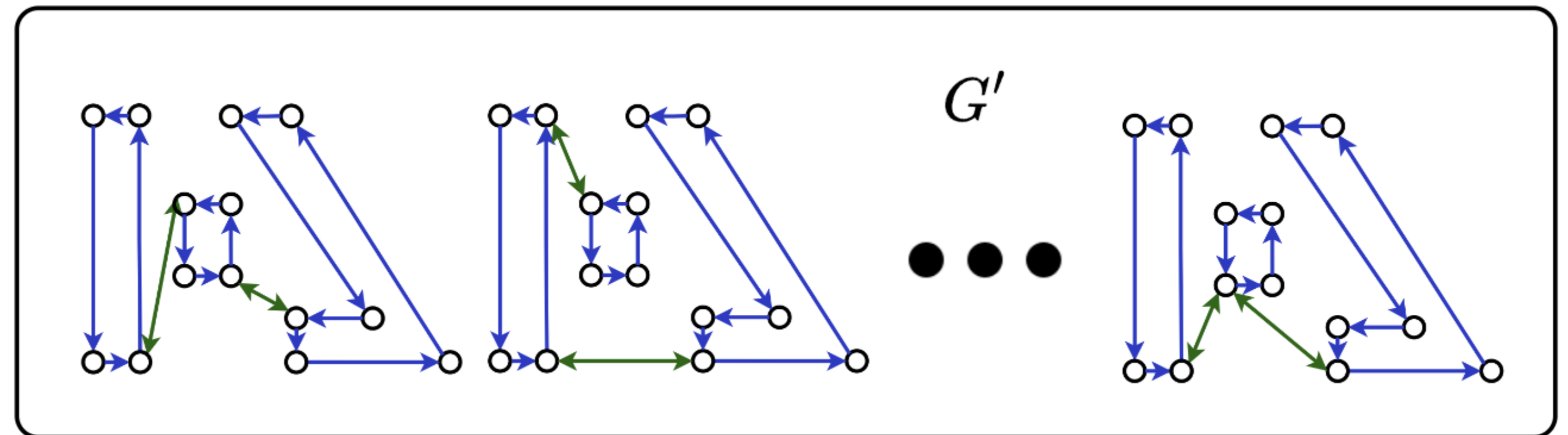
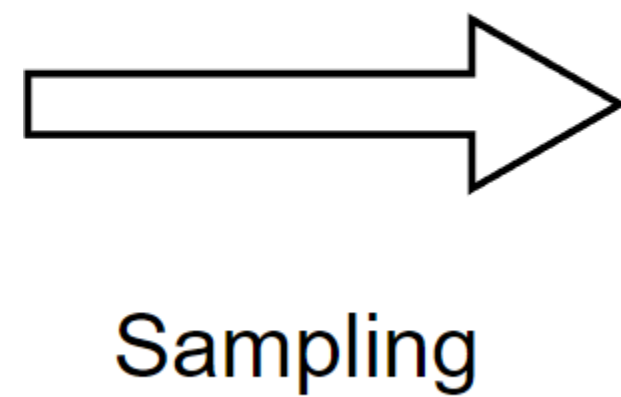
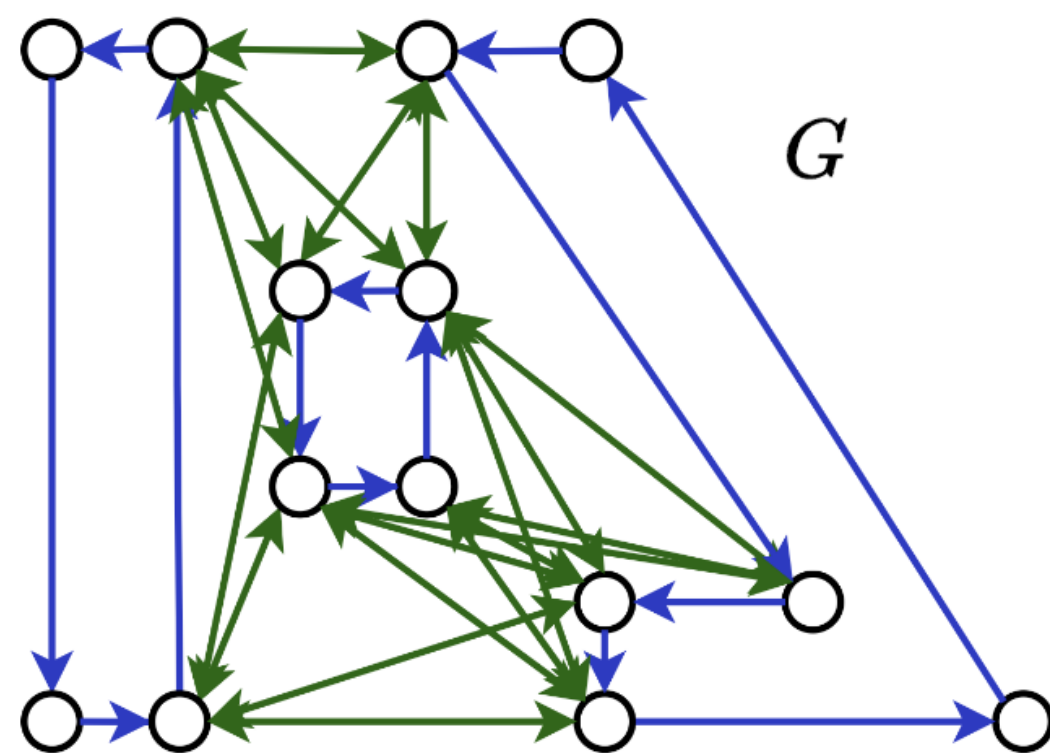


- Node  $V$ : polygon vertex
- Node feature  $X$ : coordinates
- Edge  $E$ : **polygon shape** + **visibility relationship**

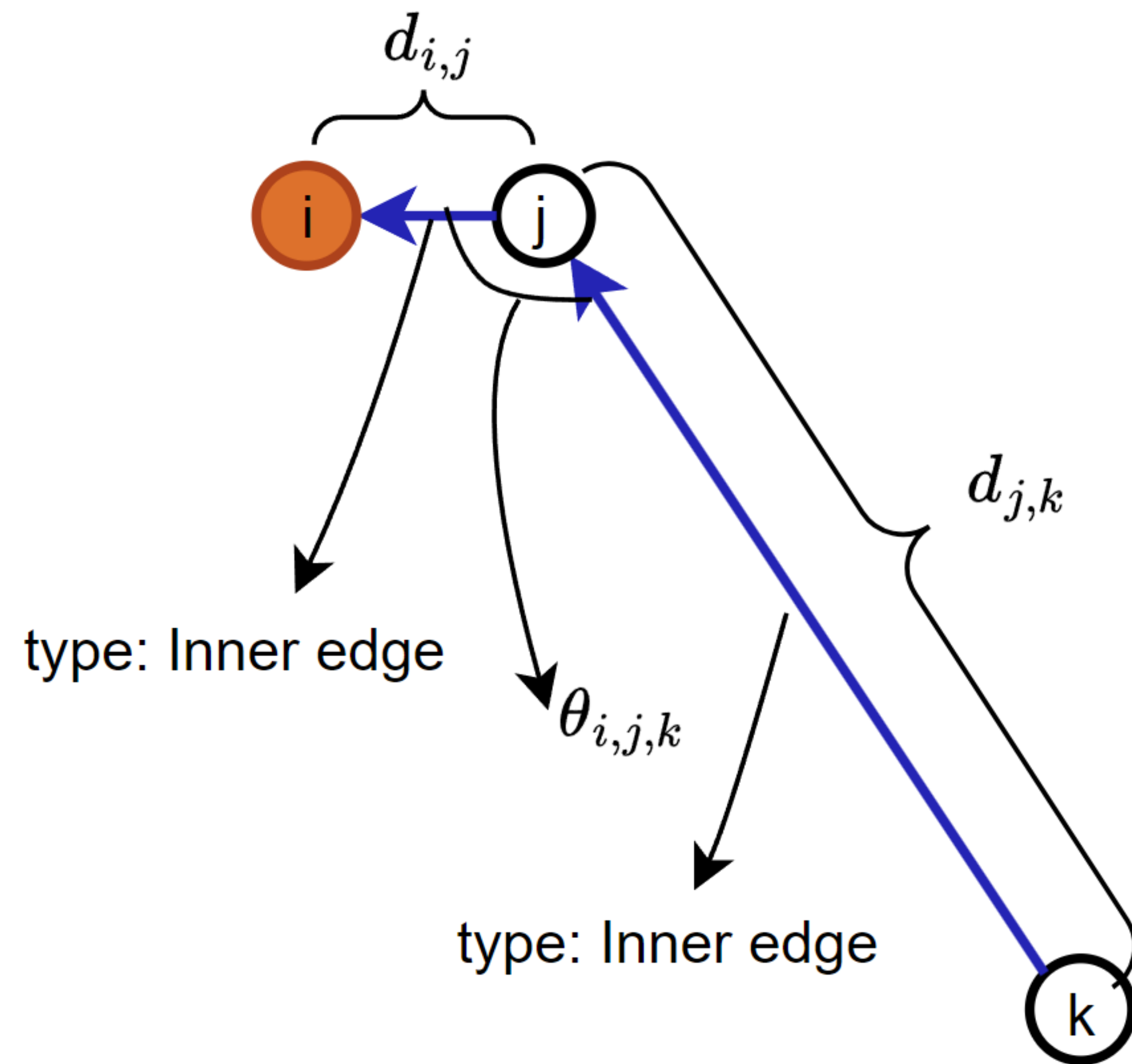


# Heterogeneous Spanning Tree Sampling

- Edge E: **polygon shape** + **visibility relationship**
- Sampling **visibility edge**

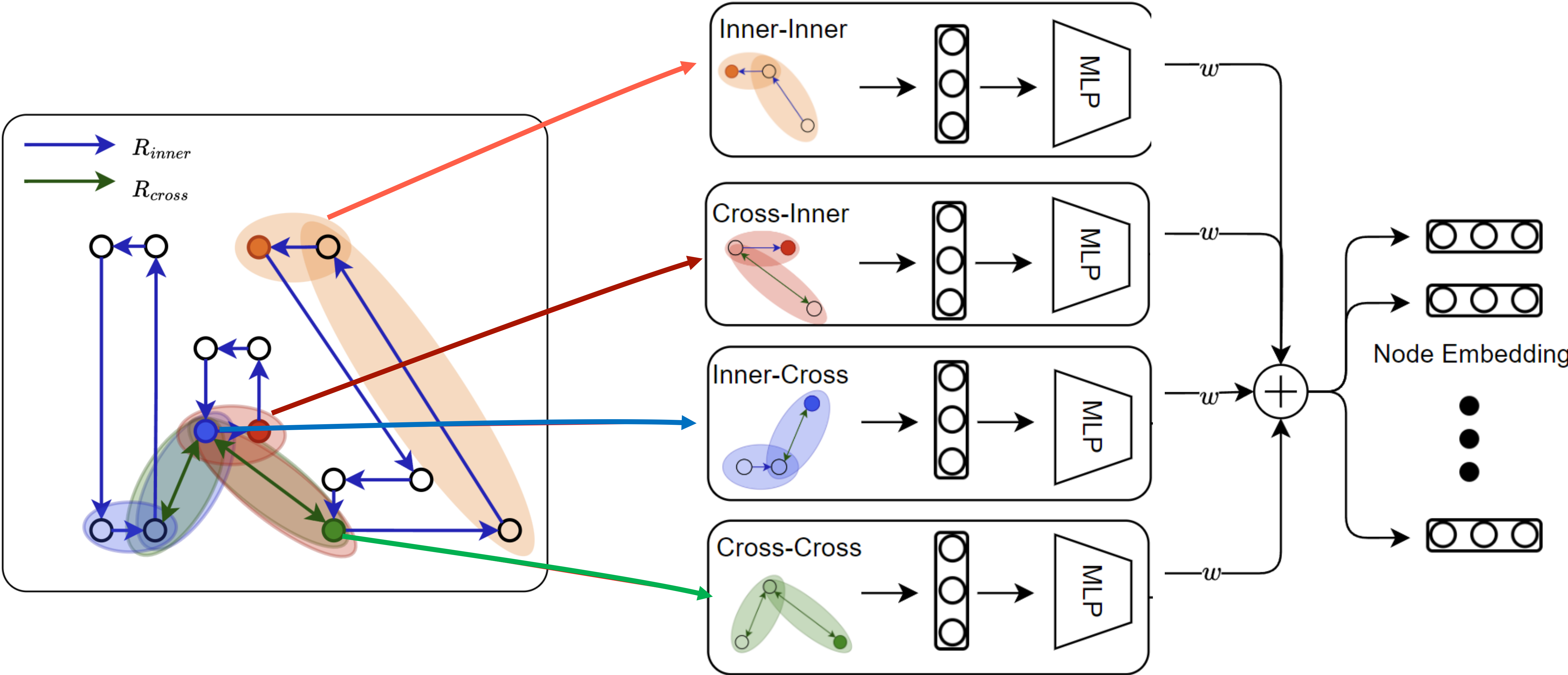


# Graph Geometric Representation



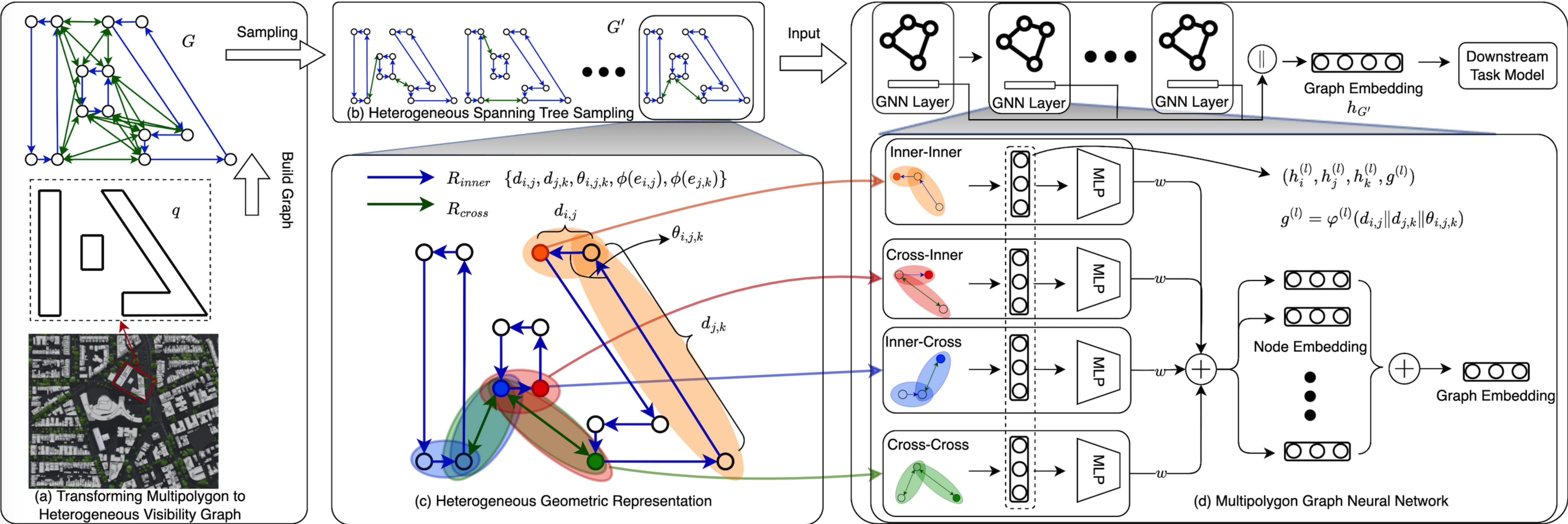
- Distance
- Distance
- Angle
- Edge Type
- Edge Type

# Multipolygon-GNN





# Overview of PolygonGNN Framework



# Datasets

Name	Number of samples	Number of classes	Polygon type	Task
MNIST-P-2	10,000	90	multipolygon	Digit classification
Building-2-R	3,469	10	multipolygon	Building shape classification
Building-2-C	5,000	10	multipolygon	Building shape classification
Building-S	5,000	10	Single polygon	Building shape classification
DBSR-cplx46K	46,567	2	multipolygon	Partially containing classification



# Effectiveness

Dataset	Metric	ResNet1D	VeerCNN	NUFT-DDSL	NUFT-IFFT	PolygonGNN
MNIST-P-2	Acc	0.794 ±.012	0.667±.019	0.559±.014	0.357±.029	<b>0.897±.004</b>
	Prec	0.810±.018	0.709±.017	0.593±.013	0.391±.027	<b>0.901±.010</b>
	F1	0.794±.012	0.667±.018	0.561±.014	0.357±.028	<b>0.897±.007</b>
	AUC	0.995±.001	0.986±.008	0.964±.005	0.908±.010	<b>0.997±.000</b>
Building-2-C	Acc	0.146±.020	0.121±.005	0.088±.023	0.059±.031	<b>0.537±.025</b>
	Prec	0.175±.026	0.125±.007	0.108±.030	0.072±.034	<b>0.578±.026</b>
	F1	0.145±.021	0.111±.006	0.086±.024	0.060±.038	<b>0.537±.025</b>
	AUC	0.860±.055	0.836±.010	0.738±.051	0.703±.055	<b>0.985±.008</b>
Building-2-R	Acc	0.464±.014	0.372±.060	0.244±.028	0.244±.013	<b>0.663±.011</b>
	Prec	0.505±.014	0.375±.111	0.278±.029	0.287±.015	<b>0.696±.021</b>
	F1	0.451±.014	0.352±.079	0.223±.023	0.229±.025	<b>0.646±.015</b>
	AUC	0.855±.005	0.843±.025	0.736±.008	0.710±.009	0.964±.008
Building-S	Acc	0.749±.016	0.643±.059	0.847±.005	0.814±.002	<b>0.984±.007</b>
	Prec	0.773±.015	0.658±.073	0.861±.005	0.846±.001	<b>0.983±.007</b>
	F1	0.748±.018	0.644±.055	0.847±.006	0.817±.001	<b>0.984±.007</b>
	AUC	0.954±.005	0.934±.021	0.986±.001	0.984±.000	<b>0.999±.000</b>
DBSR-cplx46K	Acc	0.955±.012	0.986±.001	0.990±.001	0.990±.001	<b>0.992±.001</b>
	Prec	0.956±.010	0.986±.001	0.990±.001	0.987±.001	<b>0.992±.001</b>
	F1	0.955±.012	0.986±.001	0.990±.001	0.990±.001	<b>0.992±.001</b>
	AUC	0.995±.001	0.997±.000	0.997±.000	0.997±.000	<b>0.998±.000</b>

- PolygonGNN shows the highest score among all baselines across all datasets



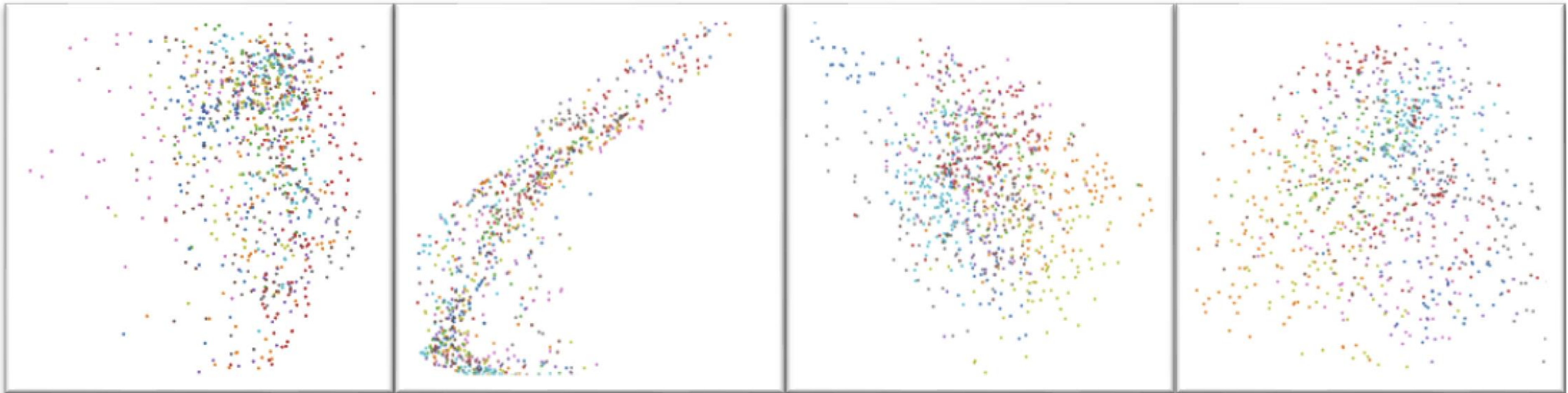
# Ablation Study

- Multipolygon-GNN
- Sampling strategy

Dataset	Metric	HAN	HGT	PolygonGNN w/o S	PolygonGNN
MNIST-P-2	Acc	0.865±.013	0.872±.016	0.880±.009	<b>0.897±.004</b>
	Prec	0.871±.012	0.877±.010	0.885±.012	<b>0.901±.010</b>
	F1	0.865±.012	0.872±.012	0.879±.010	<b>0.897±.007</b>
	AUC	0.996±.001	0.996±.001	0.996±.001	<b>0.997±.000</b>
Building-2-C	Acc	0.318±.027	0.347±.023	0.536±.013	<b>0.537±.025</b>
	Prec	0.331±.045	0.367±.041	0.568±.014	<b>0.578±.026</b>
	F1	0.310±.033	0.339±.029	0.530±.013	<b>0.537±.025</b>
	AUC	0.932±.014	0.944±.015	0.984±.003	<b>0.985±.008</b>
Building-2-R	Acc	0.599±.079	0.637±.071	0.659±.020	<b>0.663±.011</b>
	Prec	0.623±.083	0.651±.077	0.679±.020	<b>0.696±.021</b>
	F1	0.585±.080	0.625±.076	0.642±.019	<b>0.646±.015</b>
	AUC	0.917±.039	0.946±.026	<b>0.969±.010</b>	0.964±.008
Building-S	Acc	0.898±.007	0.950±.004	0.984±.009	<b>0.984±.007</b>
	Prec	0.901±.006	0.951±.004	0.984±.009	<b>0.983±.007</b>
	F1	0.898±.006	0.950±.004	0.984±.009	<b>0.984±.007</b>
	AUC	0.992±.000	0.998±.000	<b>0.999±.000</b>	<b>0.999±.000</b>
DBSR-cplx46K	Acc	0.983±.001	0.990±.002	0.990±.001	<b>0.992±.001</b>
	Prec	0.983±.001	0.990±.002	0.990±.001	<b>0.992±.001</b>
	F1	0.983±.001	0.990±.001	0.990±.001	<b>0.992±.001</b>
	AUC	0.997±.000	0.997±.000	<b>0.998±.000</b>	<b>0.998±.000</b>



# Embedding Visualization

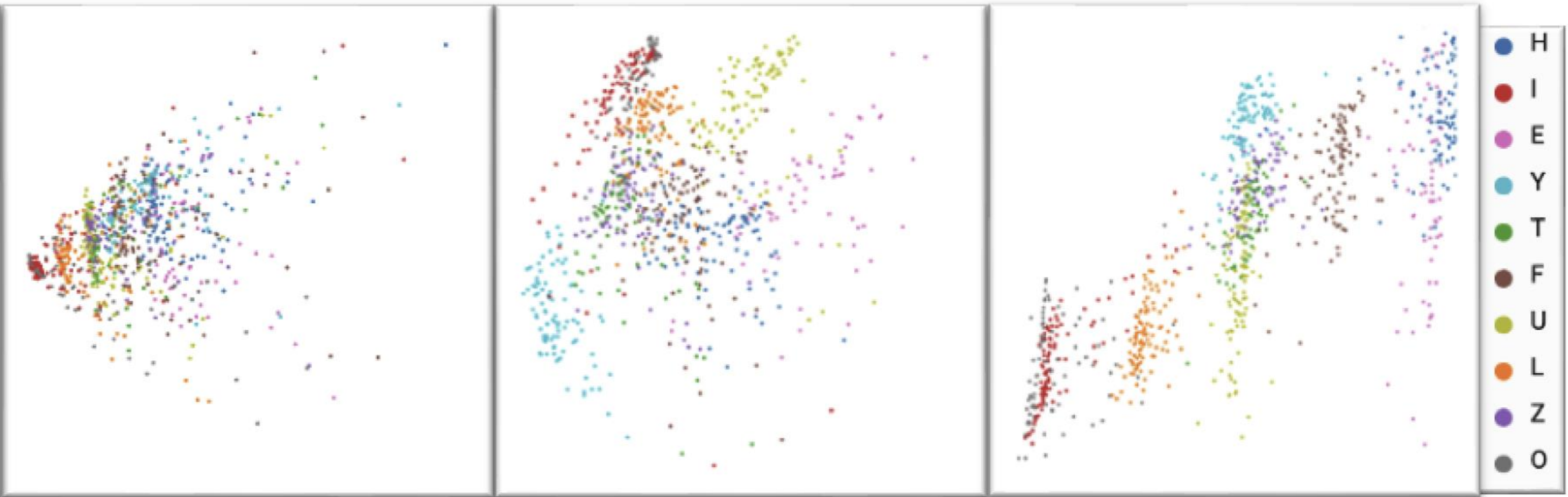


ResNet1D

VeerCNN

NUFT-DDSL

NUFT-IFFT



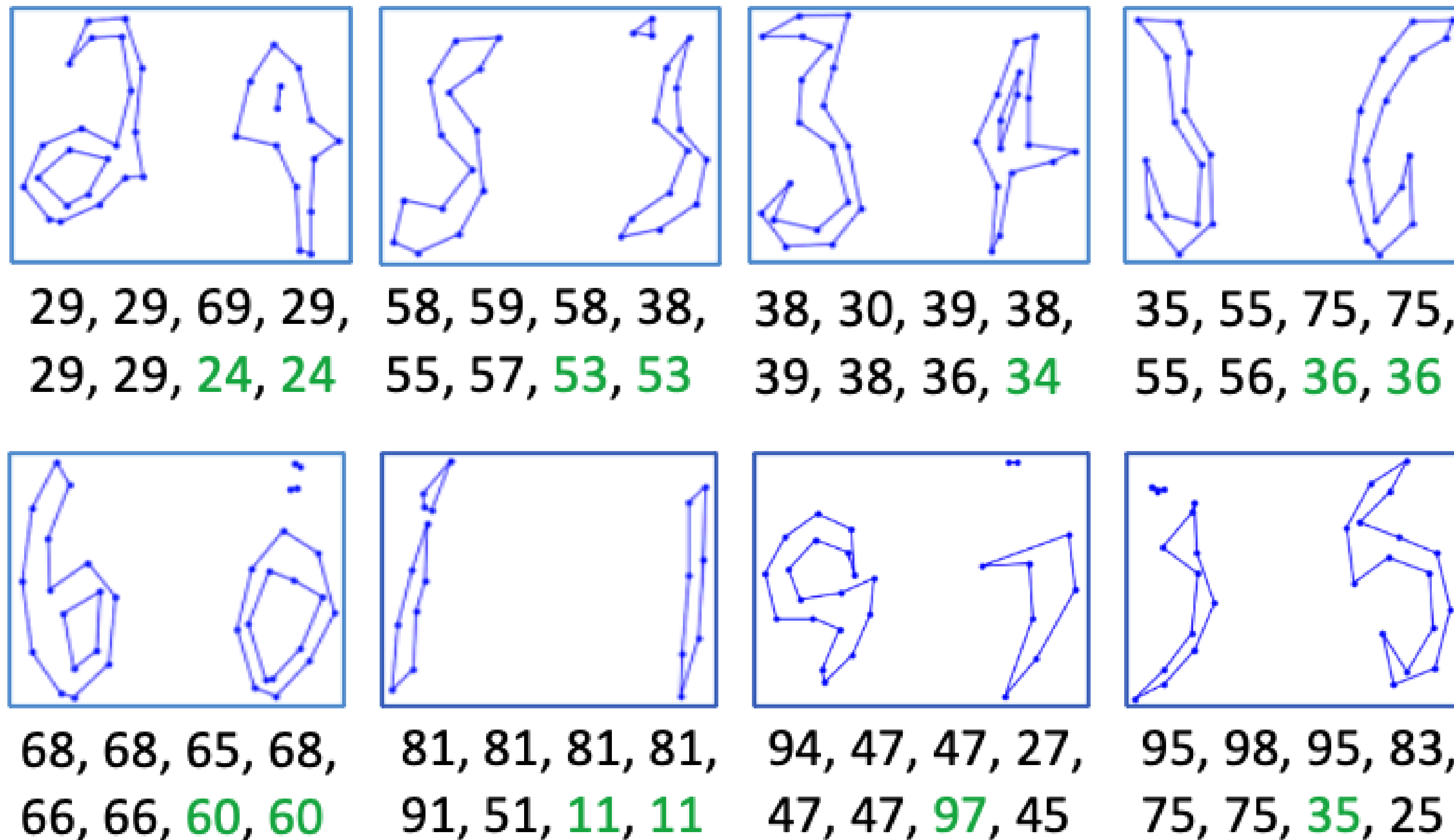
HAN

HGT

PolygonGNN



# Visualization of Prediction Cases



**Green** are  
correct  
predictions

- Predictions made by ResNet1D, VeerCNN, NUFT-DDSL, NUFT-IFFT, HAN, HGT, PolygonGNN w/o S, and PolygonGNN
- PolygonGNN adeptly handles minor imperfections

# Summary

- Heterogeneous visibility graphs conserving inner-polygonal & inter-polygonal relationships
- Heterogeneous spanning tree sampling to solve quadratic complexity from pairwise relationships
- Lossless rotation-translation-invariant geometric representation
- PolygonGNN advances the representation learning for polygonal geometries, with a particular focus on multipolygons



paper



code