

Tree Decomposed Graph Neural Network

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Tyler Derr



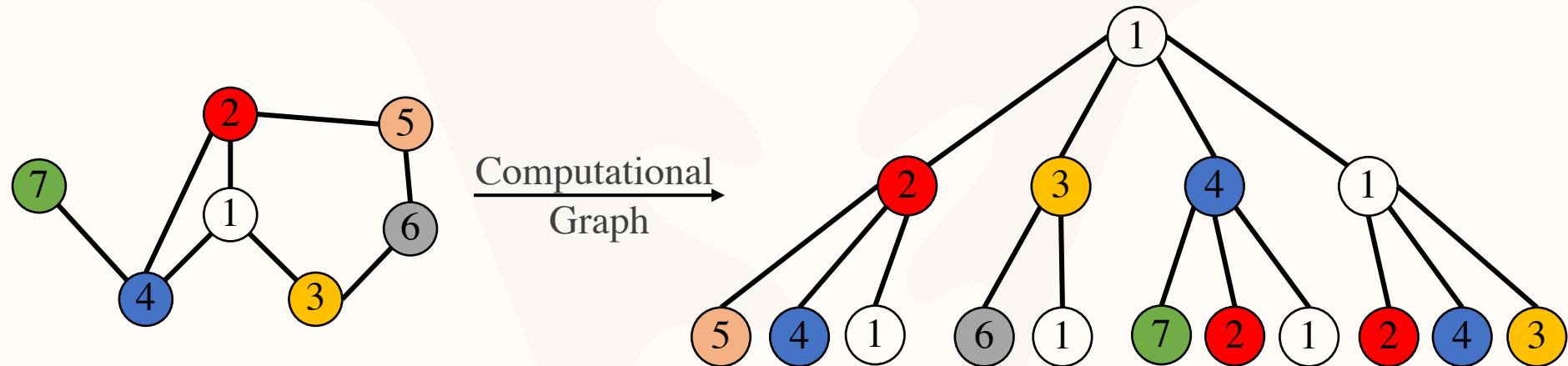
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Motivation – Tree Decomposition

Iterative propagation framework



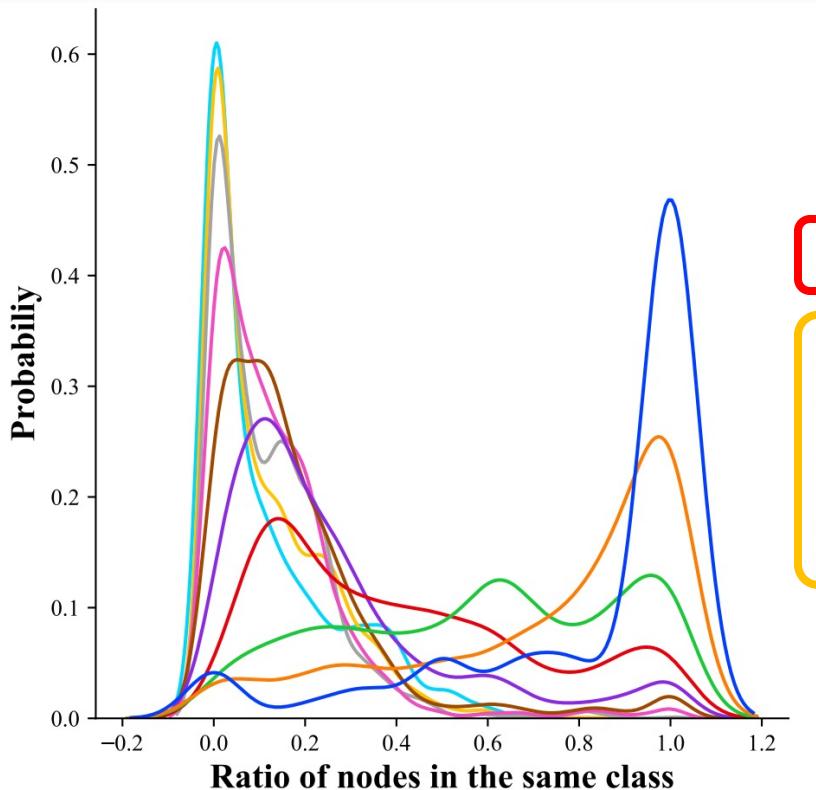
$$\hat{\mathbf{h}}_i^l = \text{AGGREGATION}^l(\mathbf{h}_i^{l-1}, \{\mathbf{h}_j^{l-1} | j \in \mathcal{N}_i\}),$$



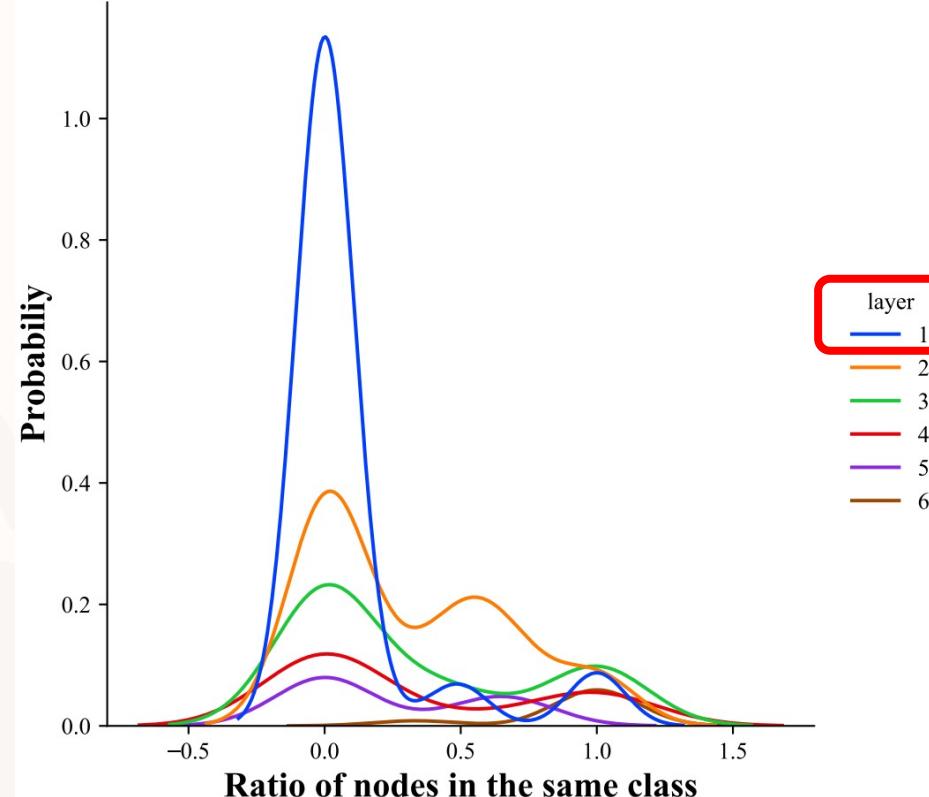
Feature smoothing between different layers!

Motivation – Tree Decomposition

Cora



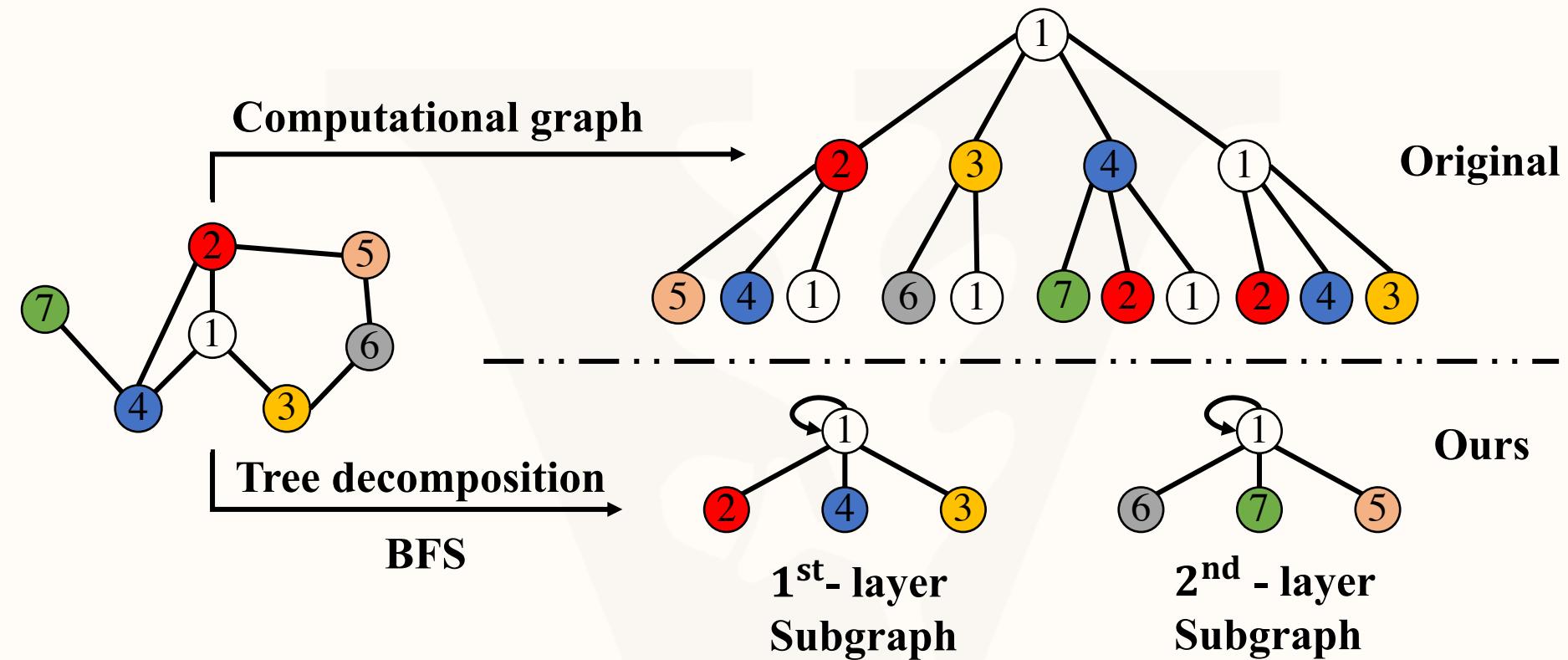
Texas



Homophily

Heterophily

Method – Tree Decomposition

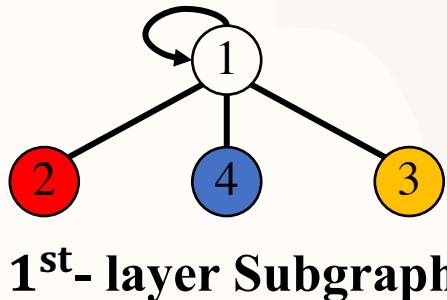


$$\mathbf{T}^k = \text{sign}(\widehat{\mathbf{A}}^k) - \text{sign}(\widehat{\mathbf{A}}^{k-1}) + \mathbf{I}$$

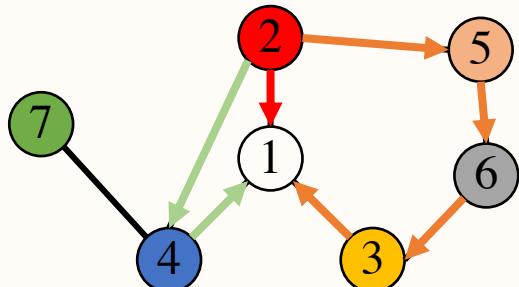
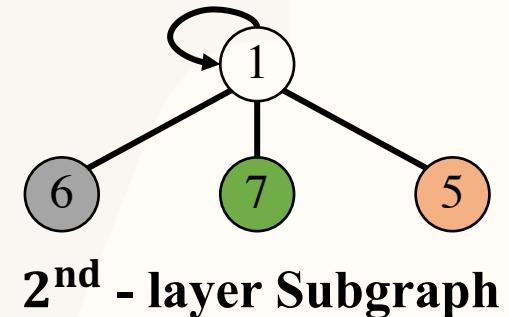
$$\text{sign}(\widehat{\mathbf{A}}^k)_{ij} = \begin{cases} 1, & \text{if } \widehat{\mathbf{A}}_{ij}^k > 0 \\ 0, & \text{if } \widehat{\mathbf{A}}_{ij}^k = 0 \end{cases}$$

Motivation – Multi-hop dependency

What's the weight for these edges?



Difference

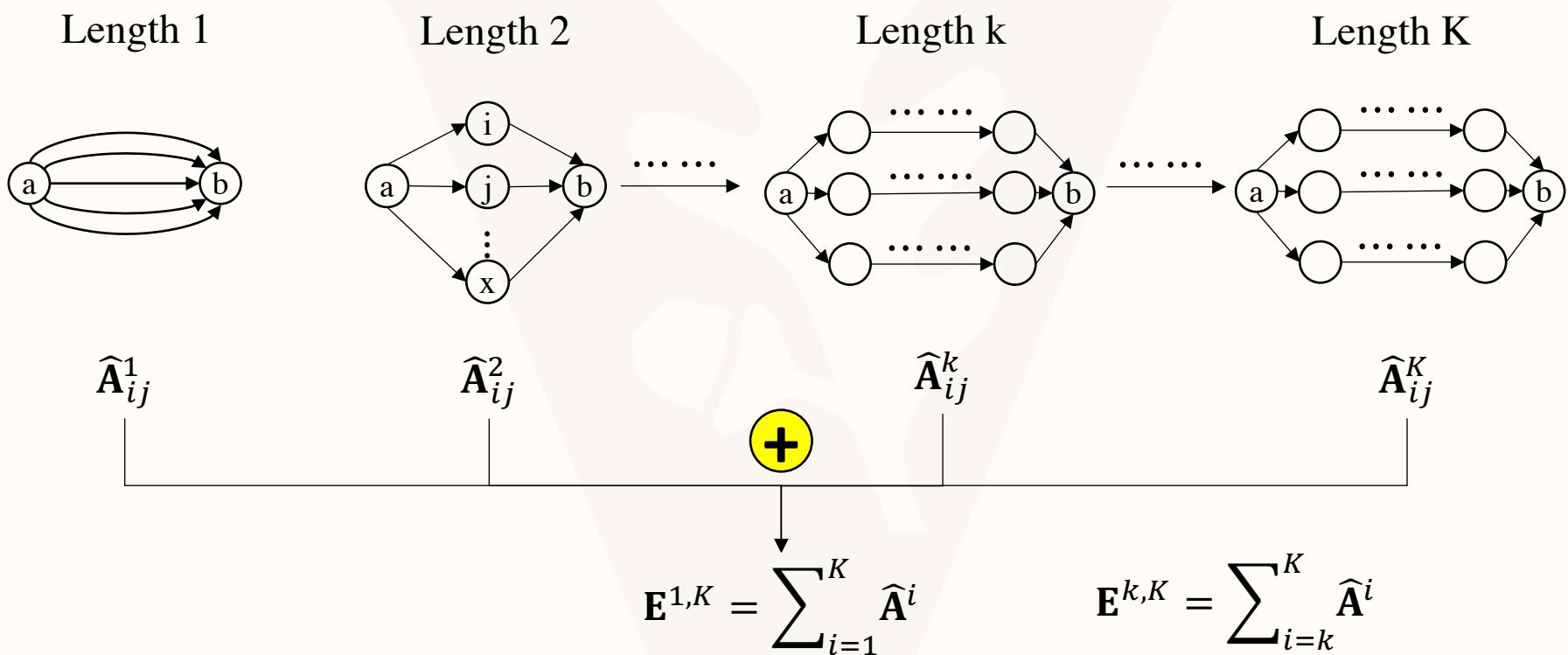


$v_2 \rightarrow v_1$
 $v_2 \rightarrow v_5 \rightarrow v_6 \rightarrow v_3 \rightarrow v_1$
 $v_2 \rightarrow v_4 \rightarrow v_1$

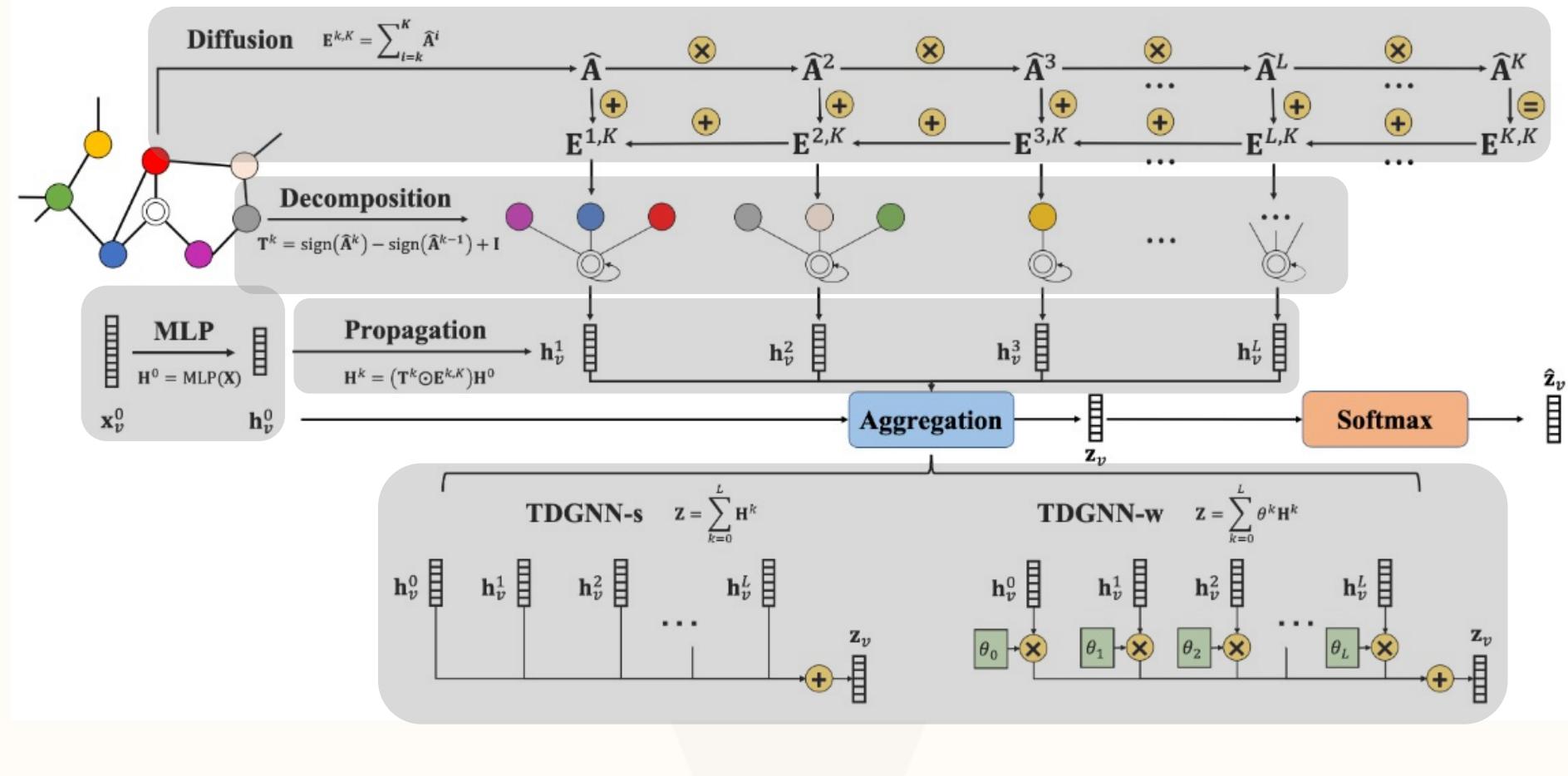
Multi-hop dependency
Width of GNNs

Method – Multi-hop dependency

Graph Diffusion



Framework - TDGNN



Experiment

Table 1: Statistics of datasets.

Networks		Nodes	Edges	Features	Classes	Train/Val/Test	Type
Homophily	Cora	2708	5429	1433	7	140/500/1000	Citation network
	Citeseer	3327	4732	3703	6	120/500/1000	Citation network
	Pubmed	19717	44338	500	3	60/500/1000	Citation network
Non-homophily	Cornell	183	295	1703	5	48%/32%/20%	Webpage network
	Texas	183	309	1703	5	48%/32%/20%	Webpage network
	Wisconsin	251	499	1703	5	48%/32%/20%	Webpage network
	Actor	7600	33544	931	5	48%/32%/20%	Actor co-occurrence network

Semi-supervised setting

- (1) 20 training nodes each class
- (2) Fixed splitting and random splitting

Full-supervised setting

- (1) 48% training nodes each class
- (2) Fixed splitting

Experiment - Semi-supervised classification

Table 2: Summary of semi-supervised classification accuracy (%) \pm stdev over Cora, Citeseer, and Pubmed datasets.

Method	Cora		Citeseer		Pubmed		Avg. Rank
	Fixed	Random	Fixed	Random	Fixed	Random	
GCN	81.50 \pm 0.79 (0-2)	79.91 \pm 1.64 (0-2)	71.42 \pm 0.48 (0-2)	68.78 \pm 2.01 (0-2)	79.12 \pm 0.46 (0-2)	77.84 \pm 2.36 (0-2)	7.17
GAT	83.10 \pm 0.40 (0-2)	80.80 \pm 1.60 (0-2)	70.80 \pm 0.50 (0-2)	68.90 \pm 1.70 (0-2)	79.10 \pm 0.40 (0-2)	77.80 \pm 2.10 (0-2)	7.00
SGC	82.63 \pm 0.01 (0-2)	80.18 \pm 1.57 (0-2)	72.10 \pm 0.14 (0-2)	69.33 \pm 1.90 (0-2)	79.12 \pm 0.10 (0-2)	76.74 \pm 2.84 (0-2)	6.83
APPNP	83.34 \pm 0.56 (0-10)	82.26 \pm 1.39 (0-10)	72.22 \pm 0.50 (0-10)	70.53 \pm 1.57 (0-10)	80.14 \pm 0.24 (0-10)	79.54 \pm 2.23 (0-10)	3.83
DAGNN	84.88 \pm 0.49 (0-10)	83.47 \pm 1.18 (0-10)	73.39 \pm 0.37 (0-9)	70.87 \pm 1.44 (0-10)	80.51\pm0.42 (0-20)	79.52 \pm 2.19 (0-20)	2.33
GCNII*	85.57\pm0.45 (0-64)	82.58 \pm 1.68 (0-64)	73.24 \pm 0.61 (0-32)	70.04 \pm 1.72 (0-10)	80.00 \pm 0.48 (0-16)	79.03 \pm 1.68 (0-16)	3.83
TDGNN-s	85.35 \pm 0.49 (0-4)	83.84\pm1.45 (0-6)	73.78\pm0.50 (0-8)	71.27\pm1.71 (0-8)	80.20 \pm 0.33 (0-5)	80.01\pm1.96 (0-5)	1.33
TDGNN-w	84.42 \pm 0.59 (0-4)	83.43 \pm 1.35 (0-6)	72.14 \pm 0.49 (0-6)	70.32 \pm 1.37 (0-6)	80.12 \pm 0.44 (0-5)	79.77 \pm 2.04 (0-5)	3.67

TDGNN-s, TDGNN-w rank 1st and 3rd

TDGNN-s, TDGNN-w utilize less layers of neighborhoods

Random setting – more robustness to data distribution

Experiment - Full-supervised classification

Table 3: Summary of full-supervised classification accuracy (%) \pm stdev over 8 datasets

Method	Cora	Cite.	Pub.	Corn.	Tex.	Wisc.	Act.	Avg. Rank
MLP	75.78 \pm 1.84 (0)	73.81 \pm 1.74 (0)	86.90 \pm 0.37 (0)	80.97 \pm 6.33 (0)	81.32 \pm 4.19 (0)	85.38 \pm 3.95 (0)	36.60 \pm 1.25 (0)	5.57
GCN	86.97 \pm 1.32 (0-2)	76.37 \pm 1.47 (0-2)	88.19 \pm 0.48 (0-2)	58.57 \pm 3.57 (0-2)	58.68 \pm 4.64 (0-2)	53.14 \pm 6.25 (0-2)	28.65 \pm 1.38 (0-2)	8.14
GAT	87.30 \pm 1.01 (0-2)	75.55 \pm 1.32 (0-2)	85.33 \pm 0.48 (0-2)	61.89 \pm 5.05 (0-2)	58.38 \pm 6.63 (0-2)	55.29 \pm 4.09 (0-2)	28.45 \pm 0.89 (0-2)	8.00
SGC	87.07 \pm 1.20 (0-2)	76.01 \pm 1.78 (0-2)	85.11 \pm 0.52 (0-2)	58.68 \pm 3.75 (0-2)	60.43 \pm 5.11 (0-2)	53.49 \pm 5.13 (0-2)	27.46 \pm 1.46 (0-2)	8.57
Geom-GCN*	85.35 \pm 1.57 (0-2)	78.02\pm1.15 (0-2)	89.95 \pm 0.47 (N/A)	60.54 \pm 3.67 (0-2)	66.76 \pm 2.72 (N/A)	64.51 \pm 3.66 (N/A)	31.63 \pm 1.15 (N/A)	5.86
APPNP	86.76 \pm 1.74 (0-10)	77.08 \pm 1.56 (0-10)	88.45 \pm 0.42 (0-10)	74.59 \pm 5.11 (0-10)	74.30 \pm 4.74 (0-10)	81.10 \pm 2.93 (0-10)	34.36 \pm 1.09 (0-10)	5.43
DAGNN	87.26 \pm 1.42 (0-10)	76.47 \pm 1.54 (0-10)	87.49 \pm 0.63 (0-20)	80.97 \pm 6.33 (0)	81.32 \pm 4.19 (0)	85.38 \pm 3.95 (0)	36.60 \pm 1.25 (0)	4.71
GCNII*	88.27\pm1.31 (0-64)	77.06 \pm 1.67 (0-64)	90.26\pm0.41 (0-64)	76.70 \pm 5.40 (0-16)	77.08 \pm 5.84 (0-32)	80.94 \pm 4.94 (0-16)	35.18 \pm 1.30 (0-64)	3.71
TDGNN-s	88.26 \pm 1.32 (0-4)	76.64 \pm 1.54 (0-8)	89.13 \pm 0.39 (0-1)	80.97 \pm 6.33 (0)	82.95 \pm 4.59 (0, 4-5)	85.47 \pm 3.88 (0, 4-5)	36.70 \pm 1.28 (0, 3-4)	2.86
TDGNN-w	88.01 \pm 1.32 (0-5)	76.58 \pm 1.40 (0-2)	89.22 \pm 0.41 (0-1)	82.92\pm6.61 (0, 2-6)	83.00\pm4.50 (0, 2)	85.57\pm3.78 (0, 3-5)	37.11\pm0.96 (0, 3-4)	2.14

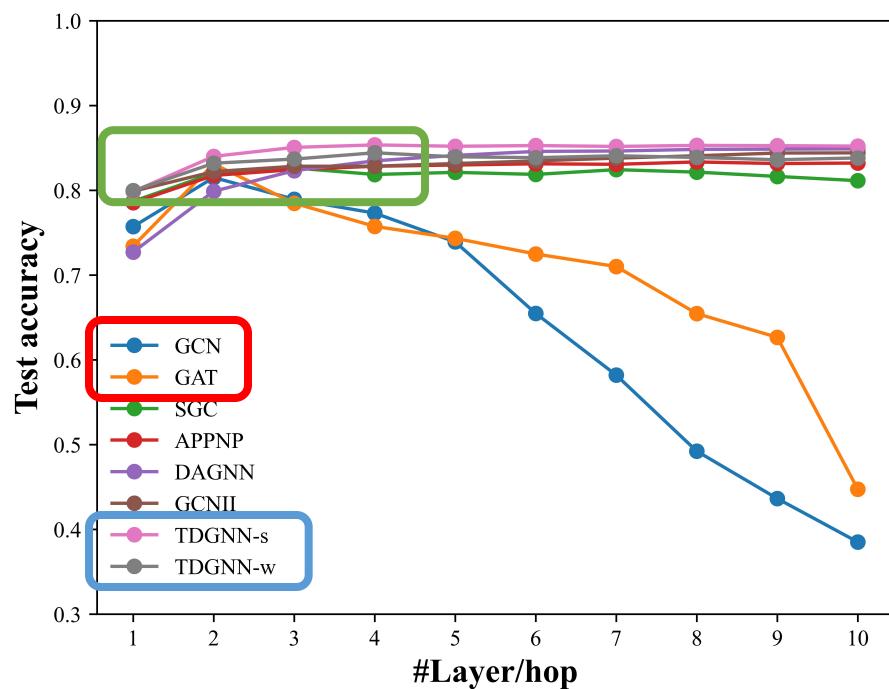
* We reuse the results reported in [33] for Geom-GCN. 'N/A' indicate the corresponding layers are not reported in the paper.

TDGNN-w, TDGNN-s rank 1st and 2nd

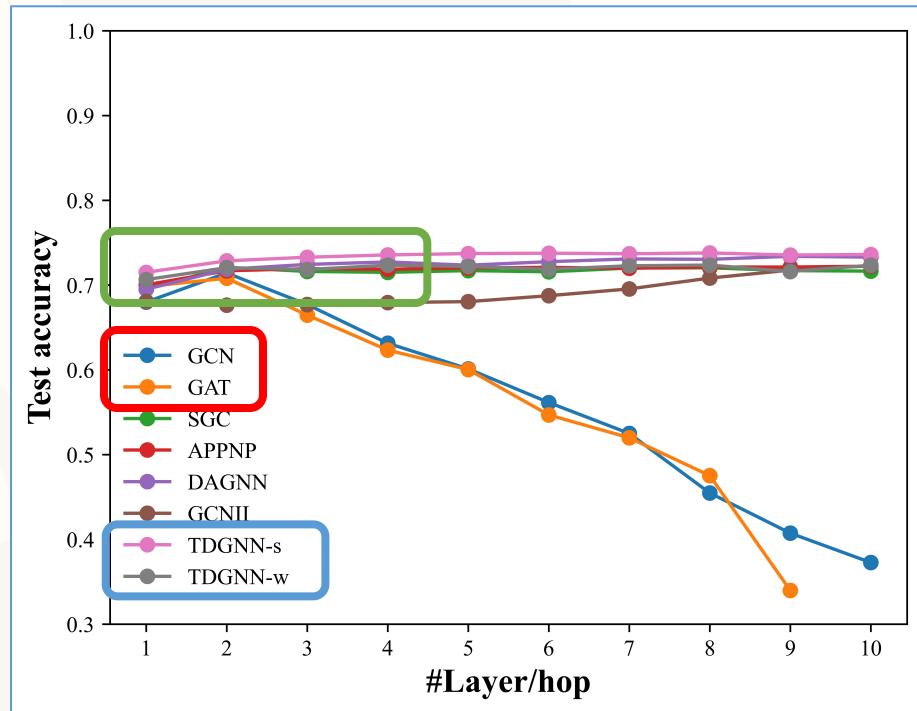
TDGNN-s, TDGNN-w leverage different layers of neighborhoods

Experiment – Further Probe

Alleviate over-smoothing

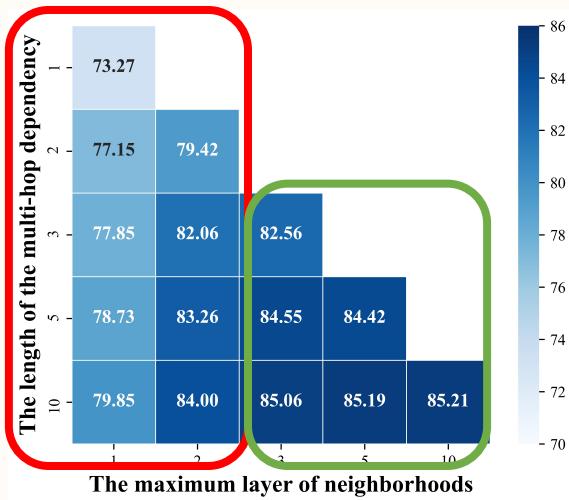


Cora

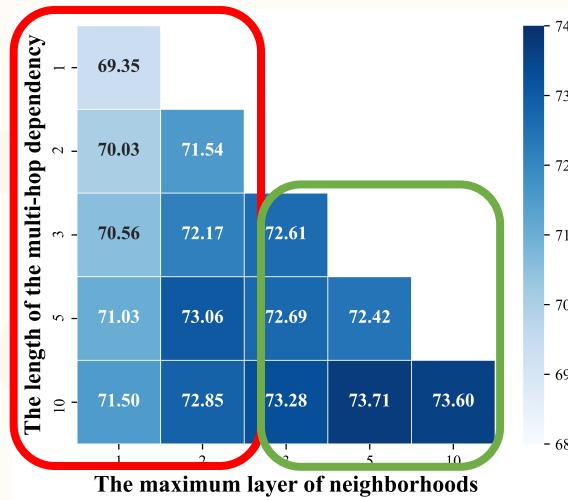


Citeseer

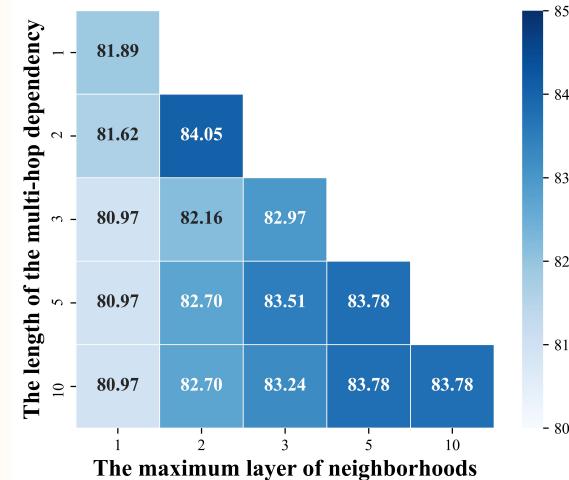
Experiment – Further Probe



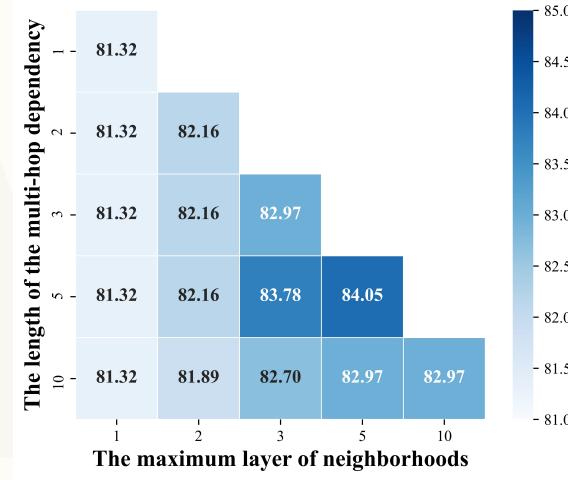
Cora



Citeseer



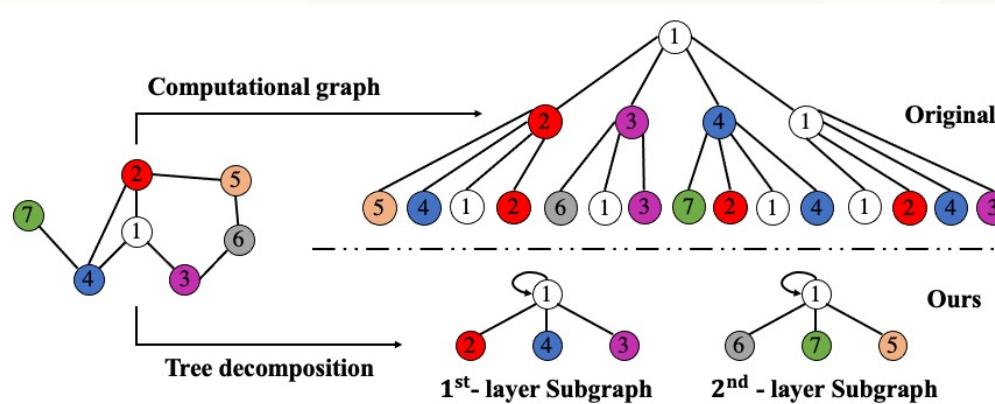
Texas



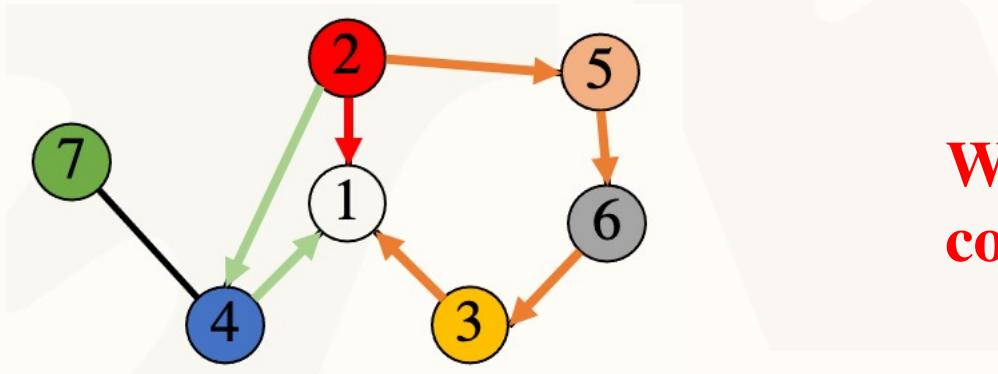
Wisconsin

Conclusion

Tree decomposition to alleviate over-smoothing between different layers



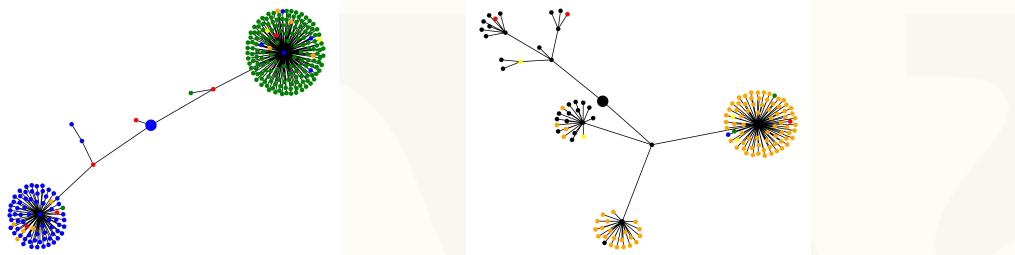
Graph diffusion to incorporate multi-hop dependencies



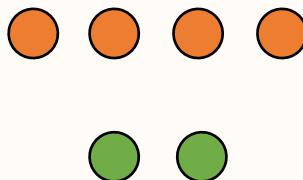
Width is also important
compared with depth!

Future Directions

Layer adaptive -> node adaptive

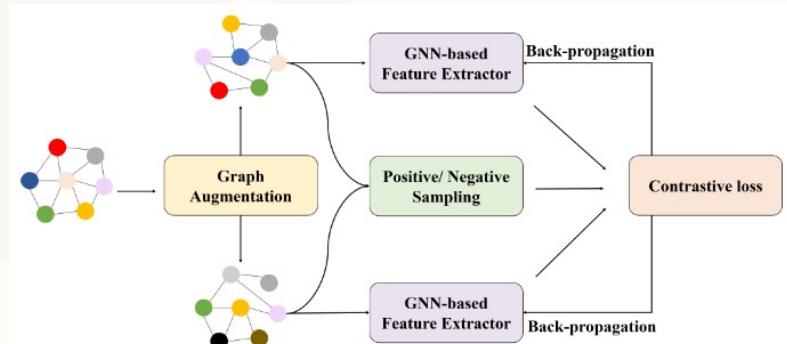
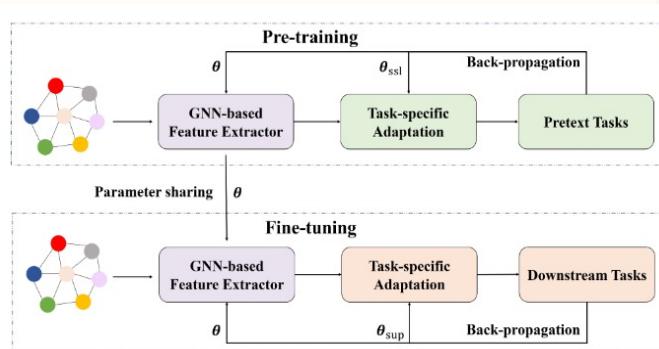


Node/Graph imbalance classification



$$\mathcal{L} = \mathcal{L}_{\min} + \mathcal{L}_{\text{maj}}$$

Incorporate self-supervised learning with deeper GNNs



Acknowledgement

Project webpage:
<https://github.com/YuWVandy/TDGNN>



NDS Lab
Network and Data Science

Summer Interns

PhD

MS

BS

High School

Thank you!
Any question?



Please see my homepage for more details!

<https://yuwvandy.github.io/>