

# ApeGNN: Node-Wise Adaptive Aggregation in GNNs for Recommendation

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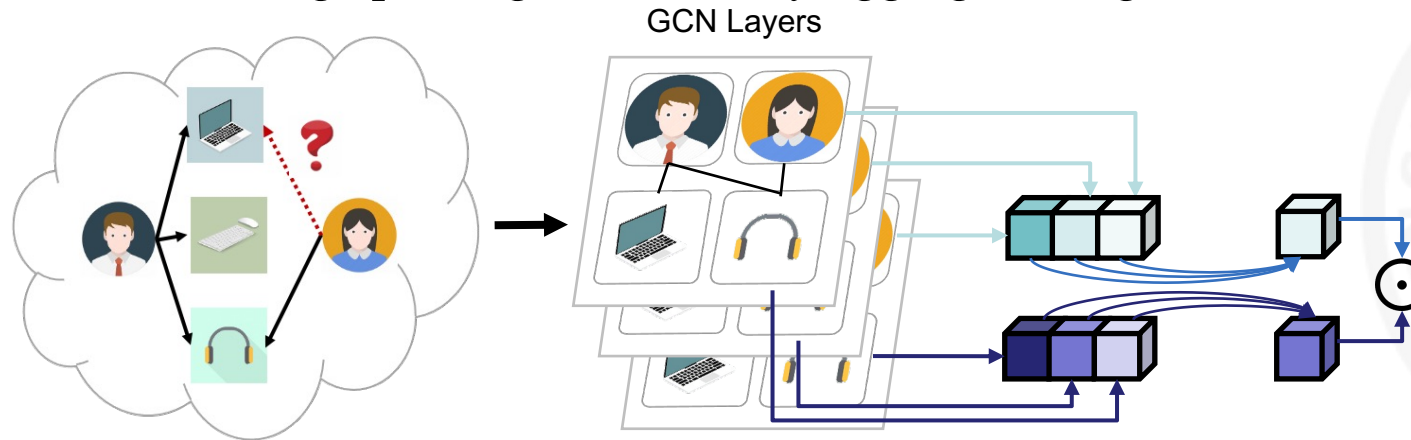
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# ApeGNN

- Background

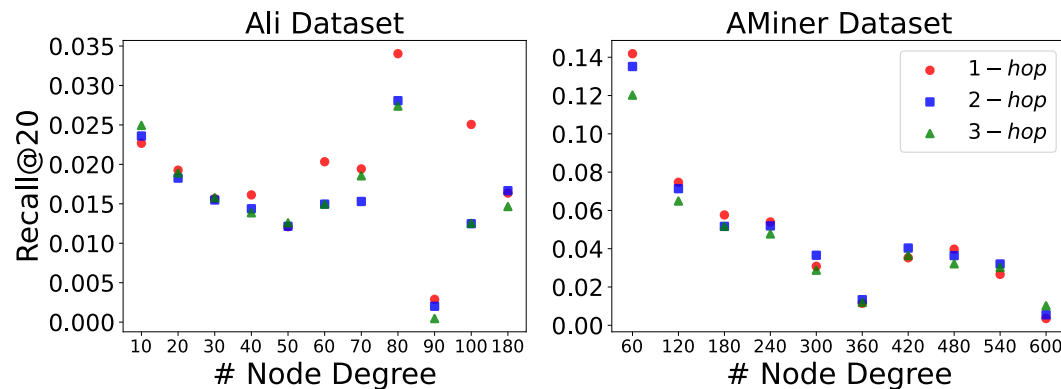
- User-item interactions can be constructed as a bipartite graph.
- Graph neural networks (GNNs) have been widely used and produced remarkable performance in recommender system.
- GNNs perform the message passing to iteratively aggregate neighborhood information.



# Why is ApeGNN?

- Motivation

- Node types are not distinguished, GNNs-based models equally treat user and item type
- Do not differentiate the local diverse pattern of each node



A motivating example. Local structures of users are diverse and node-wise aggregation is important in GNNs-based recommendation scenario.

Models	Explicit Degree Info.	Linear Propagation	Node-wise
GAT [33] <sup>[1]</sup>	✗	✗	✓
LightGCN [9] <sup>[2]</sup>	✓	✓	✗
ADC [47] <sup>[3]</sup>	✓	✓	✗
<b>ApeGNN</b>	✓	✓	✓

[1] Veličković P, Cucurull G, Casanova A, et al. Graph attention networks[J]. arXiv preprint arXiv:1710.10903, 2017.

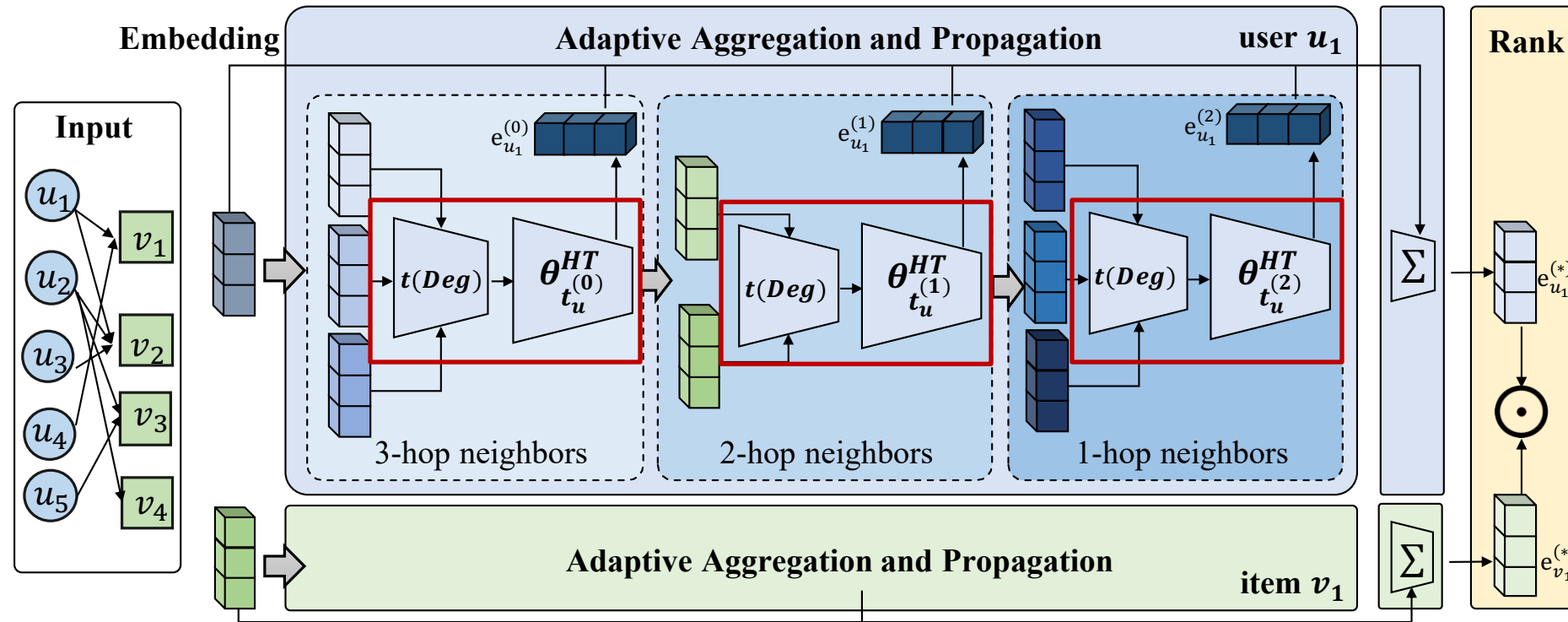
[2] He X, Deng K, Wang X, et al. Lightgcn: Simplifying and powering graph convolution network for recommendation[C]//SIGIR. 2020: 639-648.

[3] Zhao J, Dong Y, Ding M, et al. Adaptive diffusion in graph neural networks[J]. NeurIPS, 2021, 34: 23321-23333.

Node-wise Adaptive Aggregation

# What is **ApeGNN**?

## Node-wise Adaptive Aggregation



- Method
  - Develop node-wise adaptive diffusion mechanism
    - Assign unique weight for each node to all neighborhoods
    - Differentiate information of different GNN layers

# What is **ApeGNN**?

## Node-wise Adaptive Aggregation

- Diffusion mechanism

- The heat kernel.

$$\theta_{(t_{u_i}^{(l)})}^{HT} = \frac{t_{u_i}^l e^{-t_{u_i}}}{l!}.$$

- Personalized PageRank (PPR).

$$\theta_{(t_{u_i}^{(l)})}^{PPR} = t_{u_i}^{(l)} (1 - t_{u_i}^{(l)})^l.$$

- Centrality importance  $t$

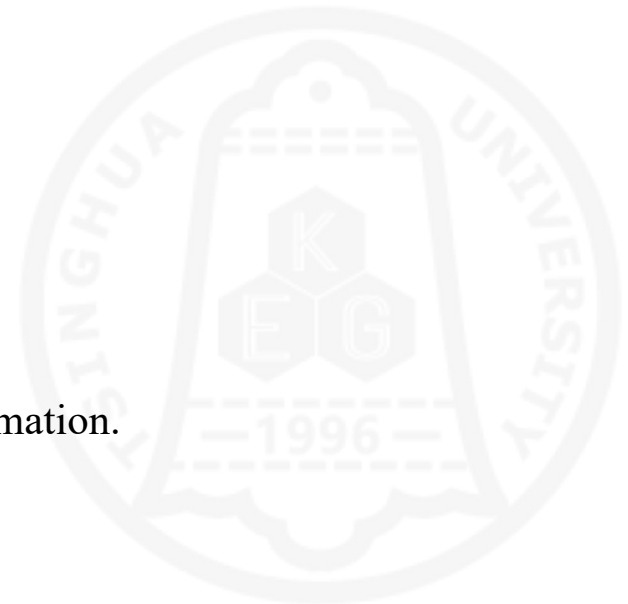
$$t_{u_i}^{(0)} = \varphi(D(u_i)) = \sigma(\log(D(u_i) + \epsilon)),$$

- Propagation process

- Adding each embedding layer to propagation layers to mine higher-order connectivity information.

- Pooling & Optimization

- Modeling the hop-wise semantic differences and optimizing the ApeGNN via BPR loss.



# How is **ApeGNN**?

## Node-wise Adaptive Aggregation

- Overall results

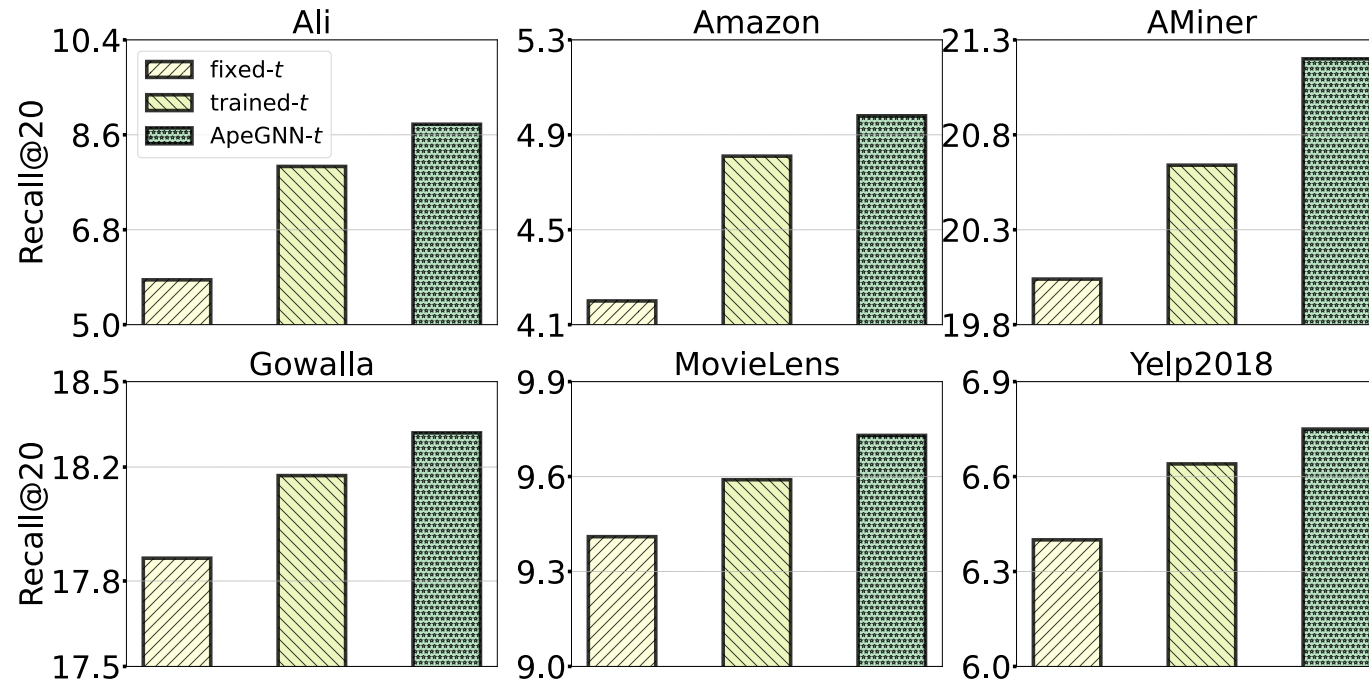
- ApeGNN outperforms SOTA GNNs-based methods on six datasets

Dataset	Ali		Amazon		AMiner		Gowalla		MovieLens		Yelp2018	
Metrics	Recall	NDCG	Recall	NDCG	Recall	NDCG	Recall	NDCG	Recall	NDCG	Recall	NDCG
BPR-MF	1.03	0.46	1.65	0.76	18.42	9.38	14.17	12.04	8.20	10.75	4.72	3.84
NeuMF	0.87	0.37	1.51	0.58	16.84	8.65	13.97	11.63	7.05	9.31	3.88	3.14
Mult-VAE	2.60	0.78	2.30	0.82	17.89	9.81	15.49	11.98	6.18	5.23	5.94	4.64
GF-CF	4.20	1.84	2.38	1.08	18.77	9.81	17.80	14.61	8.23	11.52	6.32	5.16
NGCF	4.26	1.97	2.94	1.23	17.66	9.11	14.22	11.88	9.31	11.37	5.77	4.69
LightGCN	6.13	2.86	4.11	1.86	19.69	9.87	17.75	<u>15.22</u>	9.41	11.36	<u>6.61</u>	5.39
<b>ApeGNN_HK</b>	<u>8.80</u>	<u>4.29</u>	<u>4.98</u>	<b>2.36</b>	<b>21.20</b>	<b>10.69</b>	<b>18.32</b>	<b>15.35</b>	<u>9.73</u>	<u>11.91</u>	<b>6.75</b>	<b>5.56</b>
%Improv.	43.56%	50.00%	21.17%	26.88%	7.67%	8.31%	3.21%	0.85%	3.40%	4.84	7.48%	7.96%
<b>ApeGNN_PPR</b>	<b>9.13</b>	<b>4.41</b>	<b>5.10</b>	<u>2.33</u>	<u>20.88</u>	<u>10.29</u>	<u>17.88</u>	15.17	<b>9.77</b>	<b>12.09</b>	6.59	<u>5.44</u>
%Improv.	48.94%	54.20%	24.09%	25.27%	6.04%	4.26%	0.73%	-	3.83%	6.43%	-	0.93%

# How is **ApeGNN**?

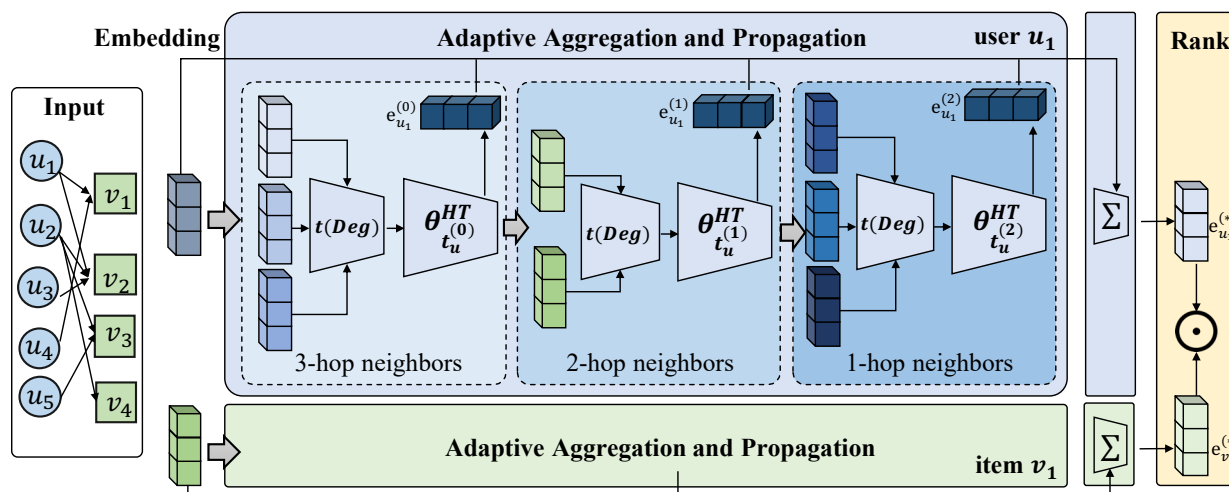
## Node-wise Adaptive Aggregation

- Ablation study
  - It demonstrates the importance and effectiveness of node-wise aggregation mechanism



# Thanks & QA!

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Dan Zhang, Yifan Zhu, Yuxiao Dong, Yuandong Wang, Wenzheng Feng, Evgeny Kharlamov, and Jie Tang. ApeGNN: Node-Wise Adaptive Aggregation in GNNs for Recommendation. In Proceedings of the th TheWebConf (WWW'23).

Code & Data <https://github.com/THUDM/ApeGNN>

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