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# BENCHTEMP: A General Benchmark for Evaluating Temporal Graph Neural Networks

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## Authors' Response to Reviewer Uv1K

**Opportunities For Improvement:** This work has a limitation in that it focuses solely on datasets with a small number of nodes. It has been acknowledged that certain Dynamic Graph Neural Networks (GNNs) struggle to handle large-scale graphs efficiently in terms of both runtime and numerical performance. Notably, the largest dataset considered in this study is Taobao, which comprises 82,566 nodes. However, real-world temporal graphs typically consist of significantly larger node counts, presenting a potential challenge for the applicability of these benchmarks w.r.t. real-world scenarios. To make the setting more realistic, you can add large-scale datasets, such as the DGraph dataset from [1] and YouTube-Reddit dataset from [2].

[1] DGraph: A Large-Scale Financial Dataset for Graph Anomaly Detection.

[2] Predicting Information Pathways Across Online Communities.

### General Response:

We appreciate your great feedback! we have included new datasets with up to several million edges and nodes. We have carefully thought your comments and added *six* datasets (eBay-Small, eBay-Large, Taobao-Large, DGraphFin, YouTubeReddit-Small, YouTubeReddit-Large), including *four large-scale* datasets (eBay-Large, Taobao-Large, DGraphFin, YouTubeReddit-Large) and corresponding experiments and detailed discussions in the updated paper. The eBay datasets are a collection of the user transactions on **eBay e-commerce platform**. We thank eBay company for sharing their datasets in our research. Considering user privacy and security, eBay datasets could only be shared among collaborators. Any researchers who are interested in the eBay datasets, please email our team. For easy access, all datasets have been hosted on the open-source platform zenodo (<https://zenodo.org/>) with a Digital Object Identifier (DOI) 10.5281/zenodo.8267846 (<https://zenodo.org/record/8267846>).

### We provide our response to each individual comment below:

#### Comment 1

This work has a limitation in that it focuses solely on datasets with a small number of nodes.

### Response:

We thank the reviewer for the strong support! We have added *six* datasets (eBay-Small, eBay-Large, Taobao-Large, DGraphFin, YouTubeReddit-Small, YouTubeReddit-Large), including *four large-scale* datasets (eBay-Large, Taobao-Large, DGraphFin, YouTubeReddit-Large). The statistics of the new datasets are shown in Table 1. For easy access, all datasets have been hosted on the open-source platform zenodo (<https://zenodo.org/>) with a Digital Object Identifier (DOI) 10.5281/zenodo.8267846 (<https://zenodo.org/record/8267846>).

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Table 1: Dataset statistics of the new datasets.

	<i>Domain</i>	<i># Nodes</i>	<i># Edges</i>
eBay-Small	E-commerce	38,427	384,677
YouTubeReddit-Small [1]	Social	264,443	297,732
eBay-Large	E-commerce	1,333,594	1,119,454
DGraphFin [2]	E-commerce	3,700,550	4,300,999
Youtube-Reddit-Large [1]	Social	5,724,111	4,228,523
Taobao-Large [3, 4]	E-commerce	1,630,453	5,008,745

- **eBay-Small** is a subset of the eBay-Large dataset. We sample 38,427 nodes and 384,677 edges from eBay-Large graph according to edge timestamps.
- **YouTubeReddit-Small** is a collection of massive visual contents on YouTube and long-term community activity on Reddit. This dataset covers a 3-month period from January to March 2020. Each row in the dataset represents a YouTube video  $v_i$  being shared in a subreddit  $s_j$  by some user  $u_k$  at time  $t$  [1]. Nodes are YouTube videos and subreddits, edges are the users’ interactions between videos and subreddits. This dynamic graph has 264,443 nodes and 297,732 edges.
- **eBay-Large** is a million-scale dataset consisting of 1.3 million nodes and 1.1 million edges, which comprises the selected transaction records from the eBay e-commerce platform over a two-month period. eBay-Large is modeled as a user-item graph, where items are heterogeneous entities which include information such as phone numbers, addresses, and email addresses associated with a transaction. We select one month of transactions as seed nodes and then expand each seed node two hops back in time to enrich the topology while maintaining consistency in the distribution of seed nodes.
- **DGraphFin** is a collection of large-scale dynamic graph datasets, consisting of interactive objects, events and labels that evolve with time. It is a directed, unweighted dynamic graph consisting of millions of nodes and edges, representing a realistic user-to-user social network in financial industry. Nodes are users, and an edge from one user to another means that the user regards the other user as the emergency contact person [2].
- **Youtube-Reddit-Large** dataset covers 54 months of YouTube video propagation history from January 2018 to June 2022 [1]. This dataset has 5,724,111 nodes and 4,228,523 edges.
- **Taobao-Large** is a collection of the Taobao user behavior dataset intercepted based on the period 8:00 to 18:00 on 26 November 2017 [4]. Nodes are users and items, and edges are behaviors between users and items, such as favor, click, purchase, and add an item to shopping cart. This public dataset has 1,630,453 nodes and 5,008,74 user-item interaction edges.

## A Experiments

We conduct extensive experiments on the tasks of *dynamic link prediction* and *dynamic node classification*. The experimental setup is the same as in the paper.

### A.1 Link Prediction Task

We run the link prediction task on 7 TGNN models and the new datasets under different settings (Transductive, Inductive, Inductive New-Old, and Inductive New-New). The AUC and AP results for each new datasets are shown in Table 2 and Table 3, respectively. For the four large-scale datasets (eBay-Large, Taobao-Large, DGraphFin, YouTubeReddit-Large), we observe the similar results as in the paper. Specifically, NAT and NeurTW achieve the top-2 performance on almost all datasets under transductive and inductive settings.

### A.2 Node Classification Task

The eBay-Small and eBay-Large datasets have node labels, so we conduct dynamic node classification experiments on both the eBay-Small and eBay-Large datasets. The AUC results are shown in Table 4.

Table 2: ROC AUC results of new datasets on the *dynamic link prediction task*. The best and second-best results are highlighted as **bold red** and underlined blue. We do not highlight the second-best if the gap is  $> 0.05$  compared with the best result.

		Transductive						
Model \ Dataset	JODIE	DyRep	TGN	TGAT	CAWN	NeurTW	NAT	
eBay-Small	0.9946 ± 0.0002	0.9941 ± 0.0006	0.9984 ± 0.0003	0.9838 ± 0.0006	0.9985 ± 0.0	<b>0.9991 ± 0.0</b>	<u>0.9978 ± 0.0003</u>	
YouTubeReddit-Small	<u>0.8519 ± 0.0007</u>	0.8499 ± 0.0012	0.8432 ± 0.0032	0.8441 ± 0.0014	0.7586 ± 0.0031	<b>0.9003 ± 0.0031</b>	0.8259 ± 0.005	
eBay-Large	0.9614 ± 0.0	0.9619 ± 0.0001	<u>0.9642 ± 0.0003</u>	0.5311 ± 0.0003	0.9442 ± 0.0003	0.9608 ± 0.0	<b>0.9658 ± 0.0002</b>	
DGraphFin	0.8165 ± 0.0024	0.8171 ± 0.0016	<b>0.8683 ± 0.0023</b>	0.6112 ± 0.0165	0.5466 ± 0.0103	<u>0.8611 ± 0.0035</u>	0.8258 ± 0.0001	
Youtube-Reddit-Large	0.8532 ± 0.0003	0.8529 ± 0.0006	0.8458 ± 0.0025	0.8536 ± 0.0026	0.7466 ± 0.0012	<b>0.916 ± 0.0025</b>	<u>0.8605 ± 0.0009</u>	
Taobao-Large	0.7726 ± 0.0005	0.7724 ± 0.001	<u>0.8464 ± 0.0008</u>	0.5567 ± 0.0047	0.7771 ± 0.0068	<b>0.859 ± 0.0091</b>	0.8188 ± 0.001	
		Inductive						
eBay-Small	0.9696 ± 0.0007	0.9674 ± 0.0018	0.9913 ± 0.0004	0.9698 ± 0.0006	0.9964 ± 0.0001	<u>0.9982 ± 0.0</u>	<b>0.9998 ± 0.0001</b>	
YouTubeReddit-Small	0.7582 ± 0.0003	0.7545 ± 0.0009	0.7276 ± 0.0033	0.7436 ± 0.0006	0.7533 ± 0.0016	0.8978 ± 0.0032	<b>0.9876 ± 0.0049</b>	
eBay-Large	0.7536 ± 0.0014	0.7515 ± 0.0006	0.7657 ± 0.0026	0.5224 ± 0.0003	0.9459 ± 0.0001	<u>0.9608 ± 0.0</u>	<b>0.9999 ± 0.0001</b>	
DGraphFin	0.6884 ± 0.0051	0.6876 ± 0.001	0.6439 ± 0.0089	0.5677 ± 0.0184	0.5479 ± 0.009	<b>0.8635 ± 0.0021</b>	<u>0.7955 ± 0.0201</u>	
Youtube-Reddit-Large	0.7539 ± 0.0005	0.7554 ± 0.0003	0.7243 ± 0.0016	0.7501 ± 0.0019	0.7327 ± 0.0016	<u>0.9128 ± 0.0031</u>	<b>0.9863 ± 0.006</b>	
Taobao-Large	0.7075 ± 0.0009	0.7042 ± 0.0006	0.6812 ± 0.0032	0.5222 ± 0.0041	0.7787 ± 0.0103	<u>0.869 ± 0.010</u>	<b>0.9933 ± 0.0008</b>	
		Inductive New-Old						
eBay-Small	0.9862 ± 0.0003	0.9836 ± 0.0016	0.9947 ± 0.0009	0.9712 ± 0.002	0.9985 ± 0.0	<u>0.9988 ± 0.0</u>	<b>0.9999 ± 0.0</b>	
YouTubeReddit-Small	0.7695 ± 0.001	0.7655 ± 0.0018	0.7396 ± 0.0034	0.7242 ± 0.0004	0.7573 ± 0.0022	<u>0.922 ± 0.0002</u>	<b>0.9967 ± 0.0014</b>	
eBay-Large	0.6109 ± 0.0244	0.5906 ± 0.0087	0.8134 ± 0.0105	0.6363 ± 0.0605	<u>0.9569 ± 0.0007</u>	0.8973 ± 0.0	<b>1.0 ± 0.0</b>	
DGraphFin	0.5768 ± 0.0071	0.5735 ± 0.0007	0.5564 ± 0.0021	0.5742 ± 0.013	0.5646 ± 0.0244	<u>0.7702 ± 0.0043</u>	<b>0.8693 ± 0.0066</b>	
Youtube-Reddit-Large	0.7844 ± 0.0015	0.7894 ± 0.0017	0.7623 ± 0.0031	0.7457 ± 0.0062	0.7511 ± 0.0022	<u>0.9356 ± 0.0004</u>	<b>0.9958 ± 0.0025</b>	
Taobao-Large	0.7023 ± 0.0015	0.6953 ± 0.0022	0.6771 ± 0.0055	0.5104 ± 0.0106	0.7674 ± 0.005	<u>0.8458 ± 0.0043</u>	<b>0.9965 ± 0.0005</b>	
		Inductive New-New						
eBay-Small	0.9388 ± 0.0009	0.9366 ± 0.0037	0.9838 ± 0.0007	0.9556 ± 0.0007	0.9937 ± 0.0	<u>0.9975 ± 0.0</u>	<b>0.9997 ± 0.0004</b>	
YouTubeReddit-Small	0.7436 ± 0.0015	0.7436 ± 0.0018	0.7265 ± 0.0055	0.749 ± 0.0011	0.7479 ± 0.004	<u>0.864 ± 0.0071</u>	<b>0.9868 ± 0.0049</b>	
eBay-Large	0.7526 ± 0.0013	0.7500 ± 0.0005	0.7639 ± 0.0027	0.5196 ± 0.0002	0.9542 ± 0.0003	<u>0.9615 ± 0.0</u>	<b>0.9999 ± 0.0001</b>	
DGraphFin	0.7307 ± 0.0007	0.7323 ± 0.0002	0.6843 ± 0.0131	0.5649 ± 0.0248	0.5417 ± 0.0099	<b>0.9051 ± 0.0028</b>	<u>0.7584 ± 0.0323</u>	
Youtube-Reddit-Large	0.6932 ± 0.0026	0.7022 ± 0.0007	0.6703 ± 0.0024	0.7269 ± 0.0	0.6942 ± 0.0028	<u>0.8716 ± 0.0077</u>	<b>0.9796 ± 0.0103</b>	
Taobao-Large	0.7243 ± 0.0001	0.7247 ± 0.0001	0.6885 ± 0.0024	0.5256 ± 0.0054	0.7922 ± 0.0118	<u>0.8906 ± 0.0088</u>	<b>0.9969 ± 0.0002</b>	

Table 3: AP results of new datasets on the *dynamic link prediction task*. The best and second-best results are highlighted as **bold red** and underlined blue. We do not highlight the second-best if the gap is  $> 0.05$  compared with the best result.

	Transductive						
Model \ Dataset	JODIE	DyRep	TGN	TGAT	CAWN	NeurTW	NAT
eBay-Small	0.9938 ± 0.0004	0.9936 ± 0.0006	<u>0.9983 ± 0.0003</u>	0.9819 ± 0.0009	0.9981 ± 0.0	<b>0.9991 ± 0.0</b>	0.9975 ± 0.0002
YouTubeReddit-Small	<u>0.8612 ± 0.0009</u>	0.8594 ± 0.0012	0.8421 ± 0.0041	0.8515 ± 0.0012	0.7625 ± 0.0042	<b>0.9112 ± 0.0021</b>	0.8325 ± 0.0068
eBay-Large	0.9318 ± 0.0002	0.9322 ± 0.0002	<u>0.9357 ± 0.0006</u>	0.5239 ± 0.0002	0.9144 ± 0.0004	0.9307 ± 0.0	<b>0.9398 ± 0.0004</b>
DGraphFin	0.7705 ± 0.0009	0.7705 ± 0.0024	<u>0.8571 ± 0.0009</u>	0.6441 ± 0.0123	0.5431 ± 0.0095	<b>0.8637 ± 0.0014</b>	0.7956 ± 0.0012
Youtube-Reddit-Large	0.8622 ± 0.0007	<u>0.8632 ± 0.0004</u>	0.8476 ± 0.0022	0.8591 ± 0.0026	0.7475 ± 0.0017	<b>0.9222 ± 0.0013</b>	0.8628 ± 0.0015
Taobao-Large	0.7164 ± 0.0003	0.7142 ± 0.0008	<u>0.844 ± 0.0011</u>	0.5761 ± 0.0023	0.7616 ± 0.0069	<b>0.8568 ± 0.016</b>	0.7904 ± 0.0008
	Inductive						
eBay-Small	0.9638 ± 0.0007	0.9619 ± 0.0017	0.9898 ± 0.0005	0.9675 ± 0.0007	0.9953 ± 0.0002	<u>0.9982 ± 0.0</u>	<b>0.9998 ± 0.0001</b>
YouTubeReddit-Small	0.7866 ± 0.0007	0.7833 ± 0.0009	0.7387 ± 0.0069	0.7551 ± 0.0002	0.7568 ± 0.0031	<u>0.9086 ± 0.0022</u>	<b>0.9872 ± 0.0056</b>
eBay-Large	0.6989 ± 0.0018	0.6973 ± 0.0007	0.7096 ± 0.0030	0.518 ± 0.0002	0.9174 ± 0.0001	<u>0.9308 ± 0.0</u>	<b>0.9999 ± 0.0001</b>
DGraphFin	0.6563 ± 0.002	0.6567 ± 0.0009	0.624 ± 0.006	0.5866 ± 0.0123	0.5428 ± 0.0082	<b>0.8626 ± 0.0012</b>	<u>0.7053 ± 0.0185</u>
Youtube-Reddit-Large	0.7796 ± 0.0009	0.7818 ± 0.0009	0.73 ± 0.0029	0.7587 ± 0.0025	0.7353 ± 0.0022	<u>0.9192 ± 0.0022</u>	<b>0.9849 ± 0.0071</b>
Taobao-Large	0.6763 ± 0.0011	0.6746 ± 0.0011	0.6664 ± 0.0012	0.5315 ± 0.0027	0.7533 ± 0.011	<u>0.8596 ± 0.0205</u>	<b>0.9941 ± 0.0007</b>
	Inductive New-Old						
eBay-Small	0.9849 ± 0.0007	0.9836 ± 0.0013	0.9931 ± 0.0008	0.9682 ± 0.0028	0.9985 ± 0.0001	<u>0.999 ± 0.0</u>	<b>0.9999 ± 0.0</b>
YouTubeReddit-Small	0.7963 ± 0.0013	0.7937 ± 0.0014	0.729 ± 0.0086	0.7296 ± 0.0013	0.762 ± 0.0041	<u>0.9244 ± 0.0015</u>	<b>0.9966 ± 0.0016</b>
eBay-Large	0.5670 ± 0.0186	0.5870 ± 0.0074	0.8024 ± 0.0060	0.6504 ± 0.0385	<u>0.9592 ± 0.0008</u>	0.8458 ± 0.0	<b>1.0 ± 0.0</b>
DGraphFin	0.6005 ± 0.0048	0.5872 ± 0.0059	0.5753 ± 0.0062	0.5927 ± 0.0058	0.5669 ± 0.0269	<u>0.7572 ± 0.0025</u>	<b>0.8184 ± 0.0088</b>
Youtube-Reddit-Large	0.808 ± 0.0014	0.8142 ± 0.0019	0.7472 ± 0.0043	0.7526 ± 0.0097	0.7553 ± 0.0025	<u>0.9368 ± 0.0009</u>	<b>0.9953 ± 0.0028</b>
Taobao-Large	0.7009 ± 0.0013	0.698 ± 0.0014	0.6879 ± 0.0008	0.5254 ± 0.0074	0.7597 ± 0.0053	<u>0.8459 ± 0.0103</u>	<b>0.9969 ± 0.0004</b>
	Inductive New-New						
eBay-Small	0.923 ± 0.001	0.9226 ± 0.0024	0.98 ± 0.0007	0.9505 ± 0.0009	0.991 ± 0.0001	<u>0.9973 ± 0.0</u>	<b>0.9997 ± 0.0004</b>
YouTubeReddit-Small	0.7578 ± 0.0015	0.7582 ± 0.0021	0.7564 ± 0.0043	0.7718 ± 0.0023	0.7498 ± 0.004	<u>0.8868 ± 0.0034</u>	<b>0.9861 ± 0.0063</b>
eBay-Large	0.6976 ± 0.0016	0.6957 ± 0.0007	0.7078 ± 0.0031	0.5154 ± 0.0001	0.93 ± 0.0003	<u>0.9318 ± 0.0</u>	<b>0.9999 ± 0.0001</b>
DGraphFin	0.6802 ± 0.0005	0.6811 ± 0.0002	0.6526 ± 0.0098	0.5831 ± 0.0184	0.5379 ± 0.0071	<b>0.8977 ± 0.0014</b>	0.6529 ± 0.0249
Youtube-Reddit-Large	0.7038 ± 0.0024	0.7115 ± 0.0007	0.6979 ± 0.002	0.7414 ± 0.0012	0.6965 ± 0.004	<u>0.8848 ± 0.0023</u>	<b>0.9761 ± 0.0134</b>
Taobao-Large	0.6738 ± 0.0005	0.6742 ± 0.0005	0.6611 ± 0.0011	0.53 ± 0.0023	0.7521 ± 0.0127	<u>0.8738 ± 0.0145</u>	<b>0.9973 ± 0.0001</b>

61 We can observe the similar results as in the paper. NeurTW achieves the best performance on both  
62 eBay-Small and eBay-Large datasets. NAT performs poorly on the node classification task.

### 63 A.3 Efficiency

Table 4: ROC AUC results for the *dynamic node classification task* on the eBay datasets. The top-2 results are highlighted as **bold red** and underlined blue.

Dataset \ Model	JODIE	DyRep	TGN	TGAT	CAWN	NeurTW	NAT
eBay-Small	0.9274 $\pm$ 0.0017	0.8677 $\pm$ 0.0356	0.913 $\pm$ 0.0025	<u>0.9342 <math>\pm</math> 0.0002</u>	0.9305 $\pm$ 0.0001	<b>0.9529 <math>\pm</math> 0.0002</b>	0.6797 $\pm$ 0.0115
eBay-Large	0.7244 $\pm$ 0.0002	0.7246 $\pm$ 0.0	0.6586 $\pm$ 0.0129	0.672 $\pm$ 0.0016	<u>0.7710 <math>\pm</math> 0.0002</u>	<b>0.7859 <math>\pm</math> 0.0</b>	0.5304 $\pm$ 0.0011

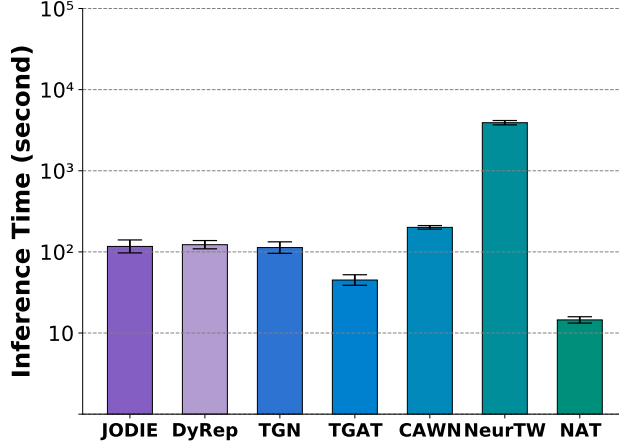


Figure 1: Inference time comparison per 100,000 edges.

64 Considering many real world applications, we add **the inference time** metric to evaluate the efficiency  
65 of methods. The inference time comparison per 100,000 edges is shown in Figure 1. According to the  
66 figure, we can observe the similar model efficiency results as in the paper. In terms of the inference  
67 time, JODIE, DyRep, TGN and TGAT are faster, while CAWN and NeurTW are much slower. NAT  
68 is relatively faster than temporal walk-based methods through caching and parallelism optimizations,  
69 *achieving a good trade-off between model quality and efficiency.*

70 **References**

- 71 [1] Yiqiao Jin, Yeon-Chang Lee, Kartik Sharma, Meng Ye, Karan Sikka, Ajay Divakaran, and Srijan Kumar.  
72 Predicting information pathways across online communities. *arXiv preprint arXiv:2306.02259*, 2023.
- 73 [2] Xuanwen Huang, Yang Yang, Yang Wang, Chunping Wang, Zhisheng Zhang, Jiarong Xu, Lei Chen, and  
74 Michalis Vazirgiannis. Dgraph: A large-scale financial dataset for graph anomaly detection. *Advances in*  
75 *Neural Information Processing Systems*, 35:22765–22777, 2022.
- 76 [3] Ming Jin, Yuan-Fang Li, and Shirui Pan. Neural temporal walks: Motif-aware representation learning on  
77 continuous-time dynamic graphs. In *Advances in Neural Information Processing Systems*, 2022.
- 78 [4] Han Zhu, Xiang Li, Pengye Zhang, Guozheng Li, Jie He, Han Li, and Kun Gai. Learning tree-based deep  
79 model for recommender systems. In *Proceedings of the 24th ACM SIGKDD International Conference on*  
80 *Knowledge Discovery and Data Mining*, pages 1079–1088, 2018.