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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 through your comments and added *six* datasets (eBay-Small, eBay-Large, Taobal-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. We provide details 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, Taobal-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.

- **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

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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

23 user  $u_k$  at time  $t$  [1]. Nodes are YouTube videos and subreddits, edges are the users’ interactions  
 24 between videos and subreddits. This dynamic graph has 264,443 nodes and 297,732 edges.

- 25 • **eBay-Large** is a million-scale dataset consisting of 1.3 million nodes and 1.1 million edges, which  
 26 comprises the selected transaction records from the eBay e-commerce platform over a two-month  
 27 period. eBay-Large is modeled as a user-item graph, where items are heterogeneous entities which  
 28 include information such as phone numbers, addresses, and email addresses associated with a  
 29 transaction. We selecte one month of transactions as seed nodes and then expand each seed node  
 30 two hops back in time to enrich the topology while maintaining consistency in the distribution of  
 31 seed nodes.
- 32 • **DGraphFin** is a collection of large-scale dynamic graph datasets, consisting of interactive objects,  
 33 events and labels that evolve with time. It is a directed, unweighted dynamic graph consisting  
 34 of millions of nodes and edges, representing a realistic user-to-user social network in financial  
 35 industry. Nodes are users, and an edge from one user to another means that the user regards the  
 36 other user as the emergency contact person [2].
- 37 • **Youtube-Reddit-Large** dataset covers **54** months of YouTube video propagation history from  
 38 January 2018 to June 2022 [1]. This dataset has 5,724,111 nodes and 4,228,523 edges.
- 39 • **Taobao-Large** is a collection of the Taobao user behavior dataset intercepted based on the period  
 40 8:00 to 18:00 on 26 November 2017 [4]. Nodes are users and items, and edges are behaviors  
 41 between users and items, such as favor, click, purchase, and add an item to shopping cart. This  
 42 public dataset has 1,630,453 nodes and 5,008,74 user-item interaction edges.

## 43 A Experiments

44 We conduct extensive experiments on the tasks of *dynamic link prediction* and *dynamic node classifi-*  
 45 *cation*. The experimental setup is the same as in the paper.

### 46 A.1 Link Prediction Task

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

### 53 A.2 Node Classification Task

54 The eBay-Small and eBay-Large datasets have node labels, so we conduct dynamic node classification  
 55 experiments on both the eBay-Small and eBay-Large datasets. The AUC results are shown in Table 4.  
 56 We can observe the similar results as in the paper. NeurTW achieves the best performance on both  
 57 eBay-Small and eBay-Large datasets. NAT performs poorly on the node classification task.

### 58 A.3 Efficiency

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	0.9642 ± 0.0003	0.5311 ± 0.0003	<u>0.9442 ± 0.0003</u>	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	0.8464 ± 0.0008	0.5567 ± 0.0047	0.7771 ± 0.0068	<b>0.859 ± 0.0091</b>	<u>0.8188 ± 0.001</u>	
		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>

59 Considering many real world applications, we add the **inference time** metric to evaluate the efficiency  
60 of methods. The inference time comparison per 100,000 edges is shown in Figure 1. According to the  
61 figure, we can observe the similar model efficiency results as in the paper. In terms of the inference  
62 time, JODIE, DyRep, TGN and TGAT are faster, while CAWN and NeurTW are much slower. NAT

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

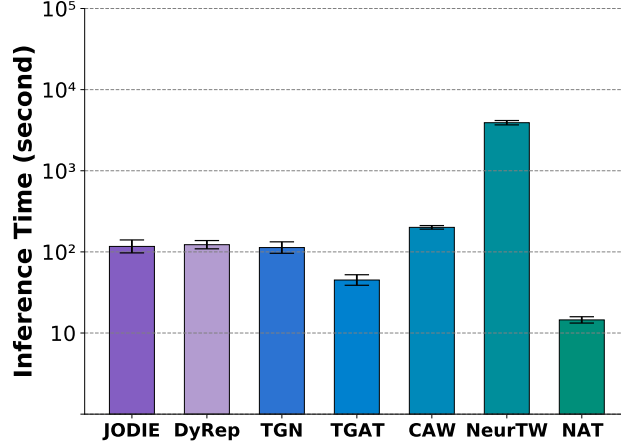


Figure 1: Inference time comparison per 100,000 edges.

63 is relatively faster than temporal walk-based methods through caching and parallelism optimizations,  
64 achieving a good trade-off between model quality and efficiency.

## 65 **References**

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