## **GLBench: A Comprehensive Benchmark** for Graph with Large Language Models

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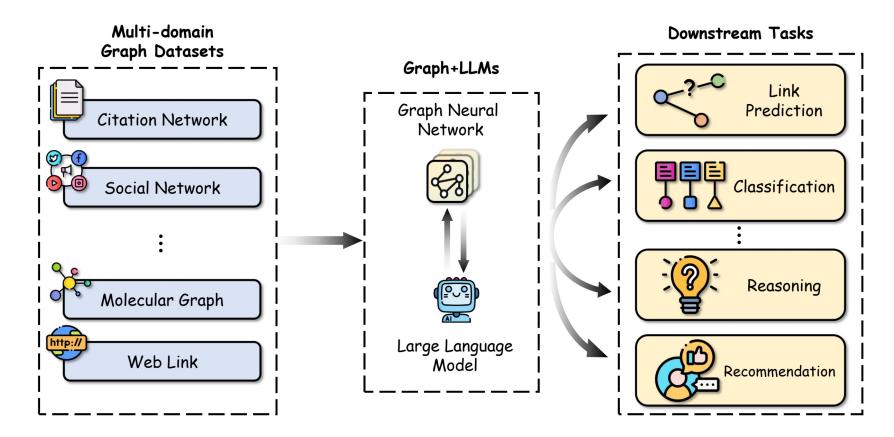
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### GraphLLM

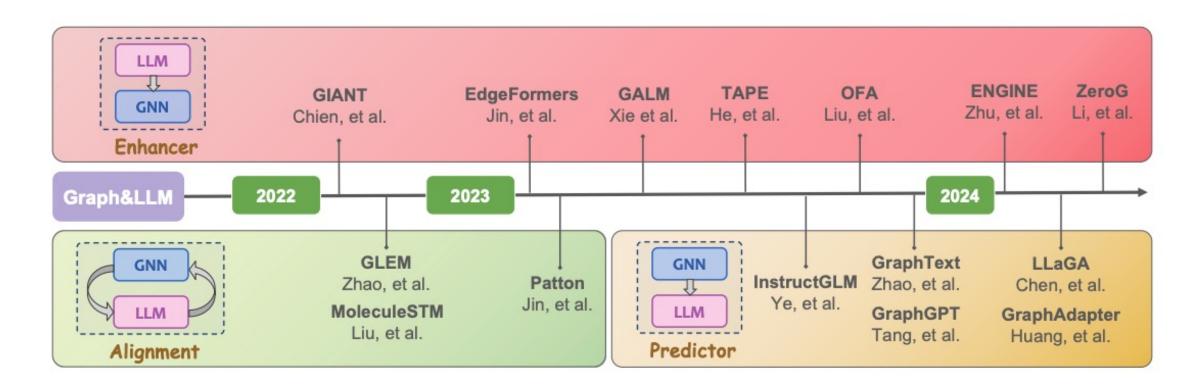
# The integration of GNNs and LLMs (GraphLLM) across a myriad of domains



Li Y, Li Z, Wang P, et al. A Survey of Graph Meets Large Language Model: Progress and Future Directions. IJCAI 2024.

### **Timeline of GraphLLM**

Existing methods can be divided into three categories based on the role palyed by LLMs.



### Benchmarking GraphLLM

Role	Method	Predictor	GNN	PLM/LLM	Techniqu	es Used	Learning	Scenarios	17	Code
Kole	Method	Fredictor	GNN	PLW/LLW	Fine-tune	Prompt	Supervised	Zero-shot	Venue	Code
	GIANT [9]	GNN	GraphSAGE, etc.	BERT	×	×	1	×	ICLR'22	Link
	TAPE [13]	GNN	RevGAT	ChatGPT	×	1	1	×	ICLR'24	Link
Enhancer	OFA [26]	GNN	R-GCN	Sentence-BERT	×	1	1	1	ICLR'24	Link
	ENGINE [54]	GNN	GraphSAGE	LLaMA-2	1	1	1	×	IJCAI'24	Link
	ZeroG [25]	GNN	SGC	Sentence-BERT	1	1	×	1	SIGKDD'24	Link
	InstructGLM [50]	LLM	-	FLAN-T5/LLaMA-v1	1	1	1	×	EACL'24	Link
	GraphText [53]	LLM	-	ChatGPT/GPT-4	1	1	1	×	Arxiv	Link
Predictor	GraphAdapter [17]	LLM	GraphSAGE	LLaMA-2	1	1	1	×	WWW'24	Link
	GraphGPT [40]	LLM	GT	Vicuna	1	1	1	1	SIGIR'24	Link
	LLaGA [6]	LLM	-	Vicuna/LLaMA-2	1	1	1	×	ICML'24	Link
Aligner	GLEM [52]	GNN/LLM	GraphSAGE, etc.	RoBERTa	1	x	1	×	ICLR'23	Link
Aligner	PATTON [21]	LLM	GT	BERT/SciBERT	1	×	✓	×	ACL'23	Link

#### Motivation

- 1. The use of different datasets, data processing approaches, and data splitting strategies in previous GraphLLM works.
- 2. The lack of benchmarks for zero-shot graph learning has led to limited exploration in this area.
- 3. Each method's computation and memory costs often overlooked.

#### Comparison with existing benchmarks

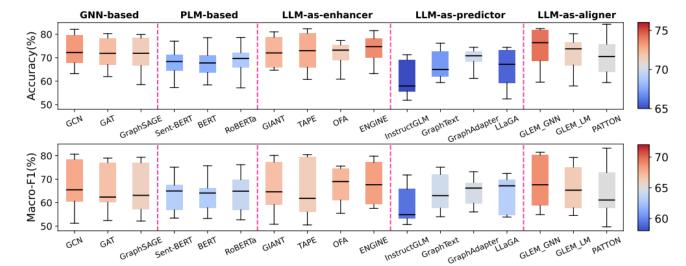
Benchmark	#Datasets (Node-level)	#Domains	Text	#Models (GraphLLM)	Model Type	Supervision Scenario
Sen et al. [38]	2 (2)	1	×	8 (0)	Classical	Supervised
Shchur et al. [39]	8 (8)	2	X	8 (0)	GNN	Supervised
OGB [15]	14 (5)	3	X	20(0)	GNN	Supervised
CS-TAG [47]	8 (6)	2	$\checkmark$	16 (2)	GNN, PLM, Enhancer	Supervised
GLBench	7 (7)	3	✓	18 (12)	GNN, PLM, GraphLLM	Supervised and Zero-shot

#### > Datasets

Dataset	# Nodes	# Edges	Avg. # Deg	Avg. # Tok	# Classes	# Train	Node Text	Domain
Cora	2,708	5,429	4.01	186.53	7	5.17%	Paper content	Citation
Citeseer	3,186	4,277	2.68	213.16	6	3.77%	Paper content	Citation
Pubmed	19,717	44,338	4.50	468.56	3	0.30%	Paper content	Citation
Ogbn-arxiv	169,343	1,166,243	13.77	243.19	40	53.70%	Paper content	Citation
WikiCS	11,701	216,123	36.94	642.04	10	4.96%	Entity description	Web link
Reddit	33,434	198,448	11.87	203.84	2	10.00%	User's post	Social
Instagram	11,339	144,010	25.40	59.25	2	10.00%	User's profile	Social

- SupervisedScenario
  - Effectiveness
  - LLM-aspredictor
  - LLM-asenhancer
  - LLM-as-aligner
  - Scaling law

Model	Co	ora	Citeseer		Pub	Pubmed		Ogbn-arxiv		WikiCS		Reddit		Instagram	
	Acc	F1	Acc	F1	Acc	F1	Acc	F1	Acc	F1	Acc	F1	Acc	F1	
GCN [23]	82.11	80.65	69.84	65.49	79.10	79.19	72.24	51.22	80.35	77.63	63.19	62.49	65.75	58.75	
GAT [43]	80.31	79.00	68.78	62.37	76.93	76.75	71.85	52.38	79.73	77.40	61.97	61.78	65.38	58.60	
GraphSAGE [10]	79.88	79.35	68.23	63.10	76.79	76.91	71.88	52.14	79.87	77.05	58.51	58.41	65.12	55.8	
Sent-BERT (22M) [36]	69.73	67.59	68.39	64.97	65.93	67.33	72.82	53.43	77.07	75.11	57.31	57.09	63.07	56.6	
BERT (110M) [22]	69.71	67.53	67.77	64.10	63.69	64.93	72.29	53.30	78.55	75.74	58.41	58.33	63.75	57.3	
RoBERTa (355M) [30]	69.68	67.33	68.19	64.90	71.25	72.19	72.94	52.70	78.67	76.16	57.17	57.10	63.57	56.8	
GIANT [9]	81.04	80.13	65.82	62.31	76.89	76.05	72.04	50.81	80.48	78.67	64.67	64.64	66.01	56.1	
TAPE [13]	80.95	79.79	66.06	61.84	79.87	79.30	72.99	51.43	<u>82.33</u>	<u>80.49</u>	60.73	60.50	65.85	50.4	
OFA [26]	75.24	74.20	73.04	68.98	75.61	75.60	73.23	57.38	77.34	74.97	<u>64.86</u>	64.95	60.85	55.4	
ENGINE [54]	<u>81.54</u>	79.82	72.15	<u>67.65</u>	74.74	75.21	<u>75.01</u>	57.55	81.19	79.08	63.20	59.34	67.62	59.2	
InstructGLM [50]	69.10	65.74	51.87	50.65	71.26	71.81	39.09	24.65	45.73	42.70	55.78	53.24	57.94	54.8	
GraphText [53]	76.21	74.51	59.43	56.43	74.64	75.11	49.47	24.76	67.35	64.55	61.86	61.46	62.64	54.0	
GraphAdapter [17]	72.85	70.66	69.57	66.21	72.75	73.19	74.45	56.04	70.85	66.49	61.21	61.13	<u>67.40</u>	<u>58.4</u>	
LLaGA [6]	74.42	72.50	55.73	54.83	52.46	68.82	72.78	53.86	73.88	70.90	67.19	67.18	62.94	54.6	
GLEM <sub>GNN</sub> [52]	82.11	80.00	71.16	67.62	81.72	<u>81.48</u>	76.43	58.07	82.40	80.54	59.60	59.41	66.10	54.9	
GLEM <sub>LLM</sub> [52]	73.79	72.00	68.78	65.32	79.18	79.25	74.03	<u>58.01</u>	80.23	78.30	57.97	57.56	65.00	54.5	
PATTON [21]	70.50	67.97	63.60	61.12	84.28	83.22	70.74	49.69	80.81	77.72	59.43	57.85	64.27	57.4	

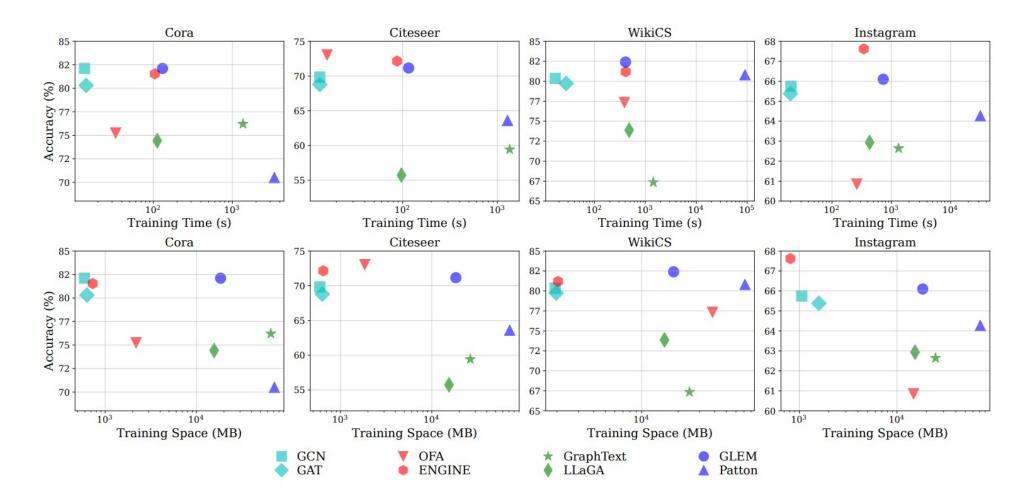


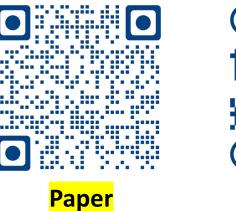
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- > Zero-shot Scenario
  - LLMs
  - Semantics/Structures?
  - □ Even a simple baseline can outperform existing GraphLLM methods.

Category	Model	$\mathcal{A}$	S	Cora		Citeseer		Pubmed		WikiCS		Instagram	
89		• •		Acc	F1	Acc	F1	Acc	F1	Acc	F1	Acc	F1
Graph SSL	DGI [44] GraphMAE [14]	√ √	X X	17.50 27.08	12.44 23.66	21.67 15.24	13.53 14.44	44.88 22.03	38.72 15.65	9.03 10.74	6.13 6.69	<b>63.64</b> 53.56	50.13 <u>52.18</u>
<u>LLMs</u>	LLaMA3 (70B) [42] GPT-3.5-turbo [35] GPT-40 [1] DeepSeek-chat [3]	× × × ×	\ \ \ \	67.99 65.67 <b>68.62</b> 65.62	68.05 63.22 68.49 65.77	51.44 50.58 <u>53.55</u> 50.35	49.98 49.34 <u>52.42</u> 48.32	77.00 75.99 77.96 <b>79.23</b>	64.18 69.90 71.79 <u>74.30</u>	<b>73.64</b> 68.75 <u>71.52</u> 70.77	<b>72.62</b> 66.56 <u>70.06</u> 69.91	38.23 49.39 42.02 40.58	36.41 49.67 40.96 39.27
Training-free	Emb w/ NA	1	1	63.59	58.23	51.75	49.51	74.66	73.15	52.30	48.40	45.52	45.14
Enhancer	OFA [26] ZeroG [25]	√ √	✓ ✓	23.11 62.52	23.30 57.53	32.45 <b>58.92</b>	28.67 <b>54.58</b>	46.60 <u>79.08</u>	35.04 <b>77.94</b>	34.27 60.46	33.72 57.24	53.63 <u>56.13</u>	51.10 <b>52.50</b>
Predictor	GraphGPT [40]	1	1	24.90	7.98	13.95	13.89	39.85	20.07	38.02	29.46	43.94	43.49

#### Efficiency









### The End, Thanks!

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