



rLLM: Relational Table Learning with LLMs

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Outline

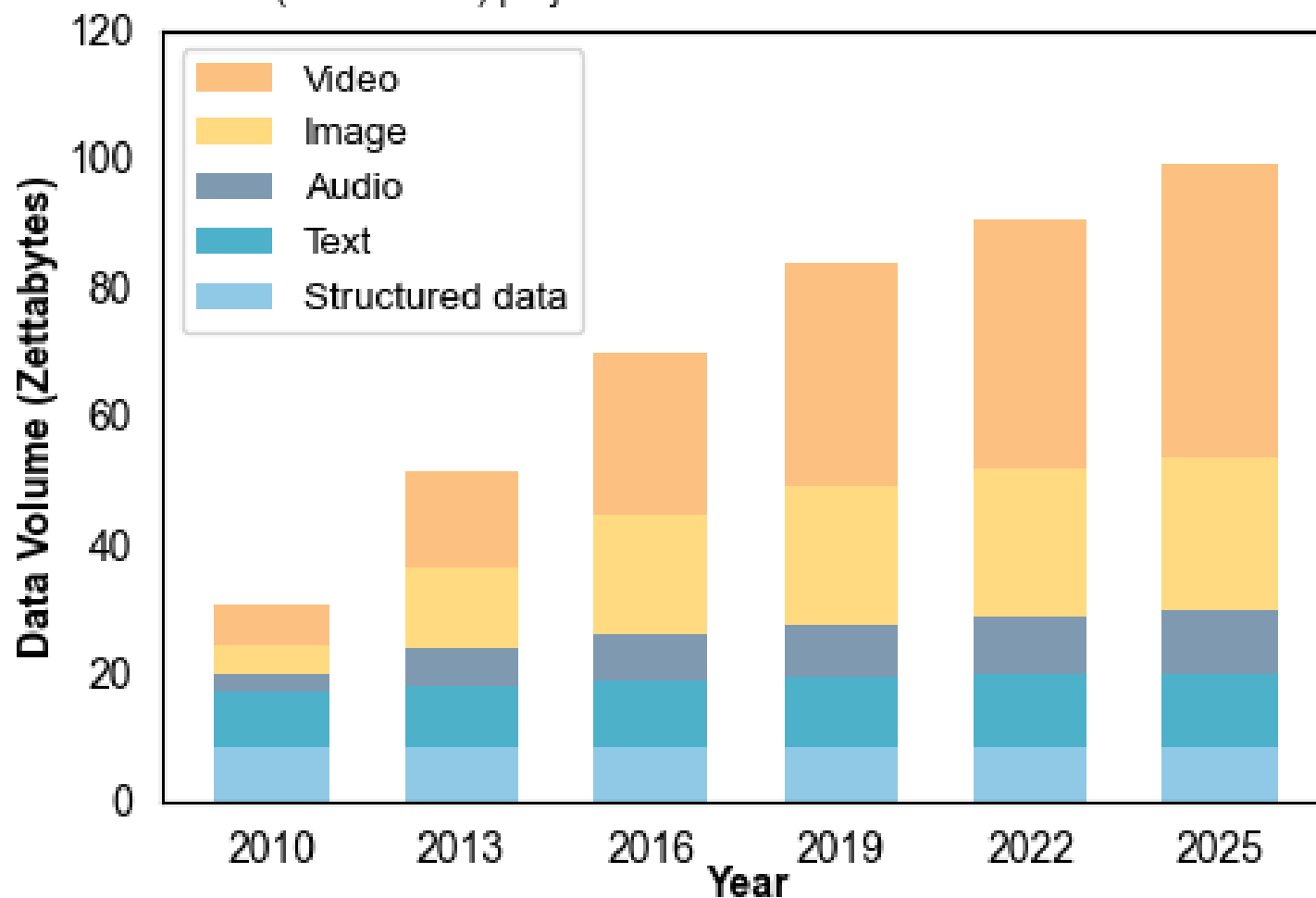
- **Background**
- rLLM (relationLLM)
- Use case

Big Data Trends



Trends in Global Data Volume by Data Type (2010-2025)

Source: The rLLM (relationLLM) project.

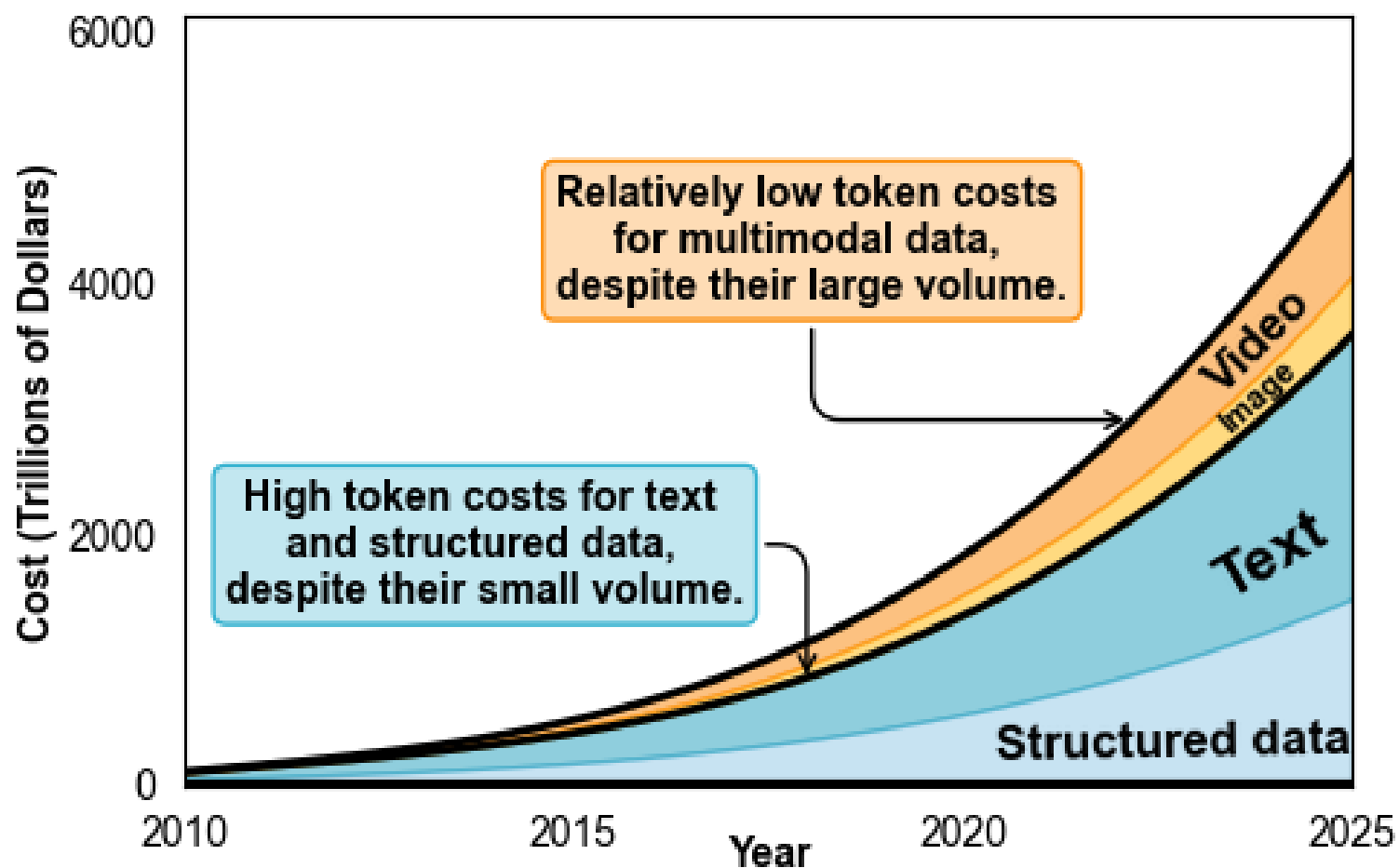


LLMs are Costly



Trends in LLM Token Costs by Data Type (2010-2025)

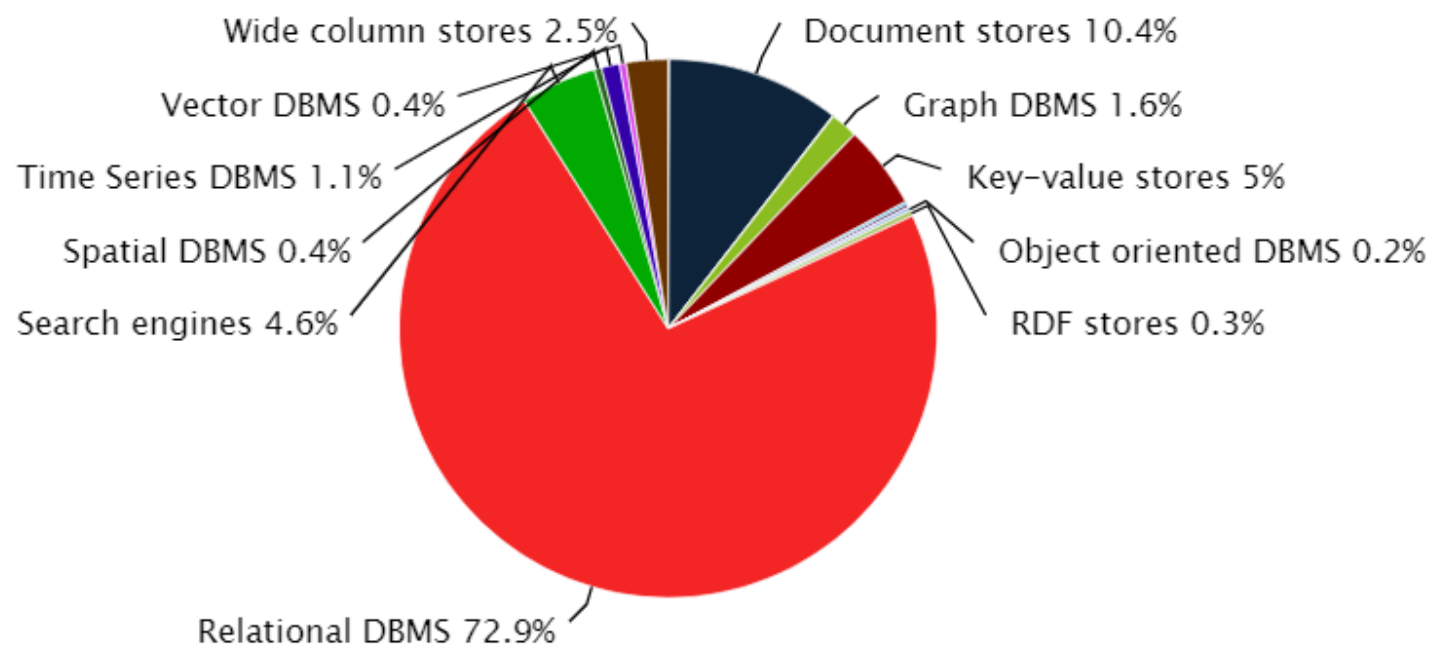
Source: The rLLM (relationLLM) project.



Relational databases domain the world data



Ranking scores per category in percent, July 2024



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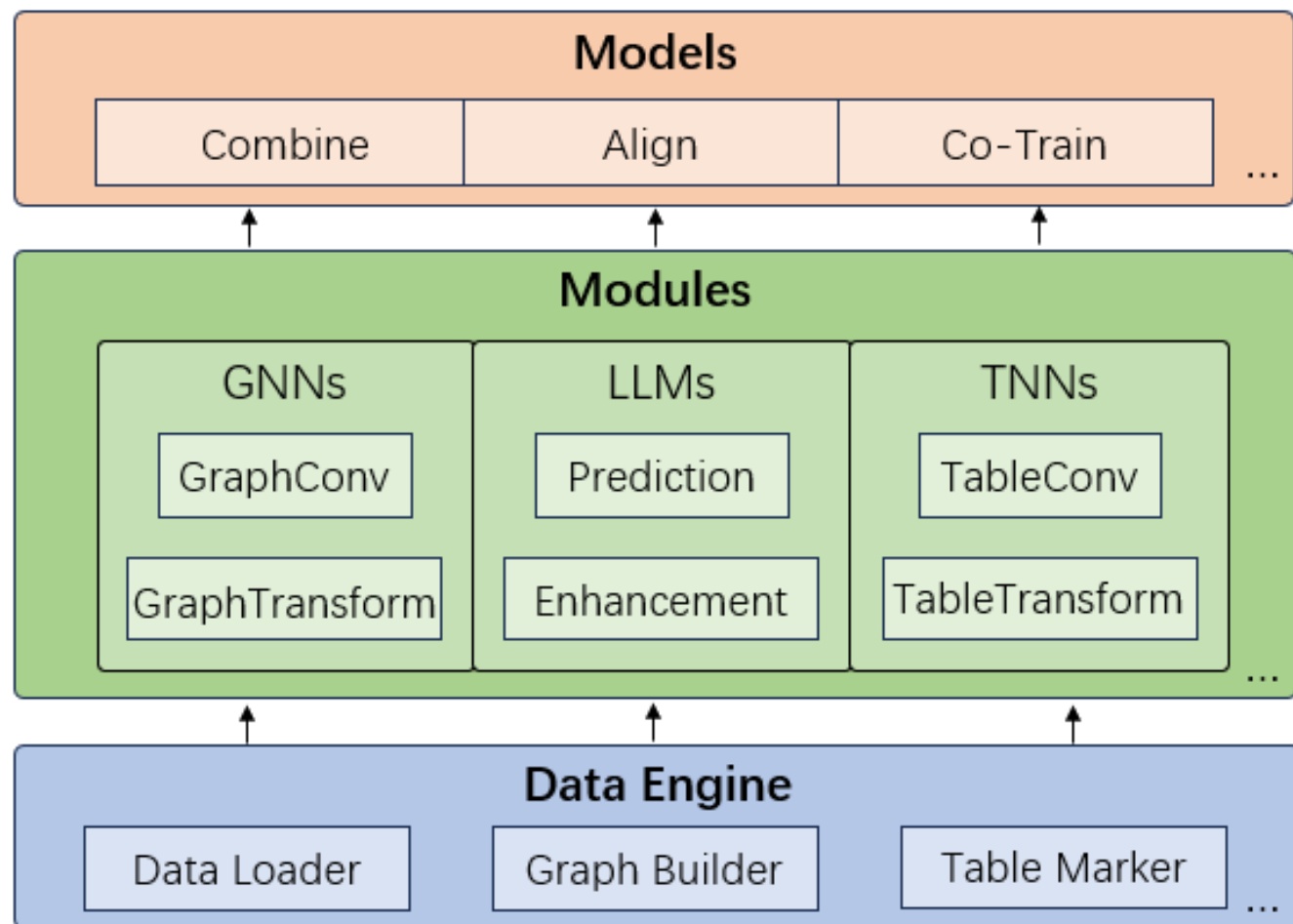
In the area of LLM, we need to pay great attention to **Relational Table Learning (RTL)**.



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- **rLLM (relationLLM)**
- Use case

Overview of rLLM



The rLLM Overall Architecture

The Key of rLLM (relationLLM)



rLLM: Relational Table Learning with LLMs

Table Learning
(for every tables)

Graph Learning
(for foreigner
keys)

LLMs
(for LLM-based
learning)

rLLM (relationLLM) is an easy-to-use Pytorch library for Relational Table Learning with LLMs, by performing two key functions:

- Breaks down state-of-the-art GNNs, LLMs, and TNNs as standardized modules.
- Facilitates novel RTL model building in a "combine, align, and co-train" way.

Project page: <https://github.com/rllm-project/rllm>

Highlight Features:



- **LLM-friendly:** Modular interface designed for LLM-oriented applications, integrating smoothly with LangChain and Hugging Face transformers.
- **One-Fit-All Potential:** Processes various graphs (like social/communication/e-commerce graphs) by treating them as multiple tables linked by foreigner keys.
- **Novel Datasets:** Introduces three new relational table datasets useful for RTL model design. Includes the standard classification task, with examples.
- **Community Support:** Maintained by students and teachers from Shanghai Jiao Tong University and Tsinghua University. Supports the SJTU undergraduate course "Content Understanding (NIS4301)" and the graduate course "Social Network Analysis (NIS8023)".



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Three new relational table datasets with standard classification tasks

- TML1M is derived from the classical MovieLens 1M dataset.
- TLF2K is derived from the classical LastFM2K dataset.
- TACM12K is derived from the ACM heterogeneous graph dataset.

Dataset	Tables [#row/#col]	Relation Tables	Label	Classes	#Train/#Val/#Test
TML1M	users [6,040/5] movies [3,883/11] ratings [1,000,209/4]	ratings: user-movie	Age range of user	7	[140/500/1000]
TLF2K	artists [9,047/10] user_artists [80,009/3] user_friends [12,717/3]	user_artists: user-artist user_friends: user-user	Genre of artist	11	[220/500/1000]
TACM12K	papers [12,499/5] authors [17,431/3] citations [30,789/2] writings [37,055/2]	citations: paper-paper writings: paper-author	Conference of paper	14	[280/500/1000]

An illustration RTL method - BRIDGE

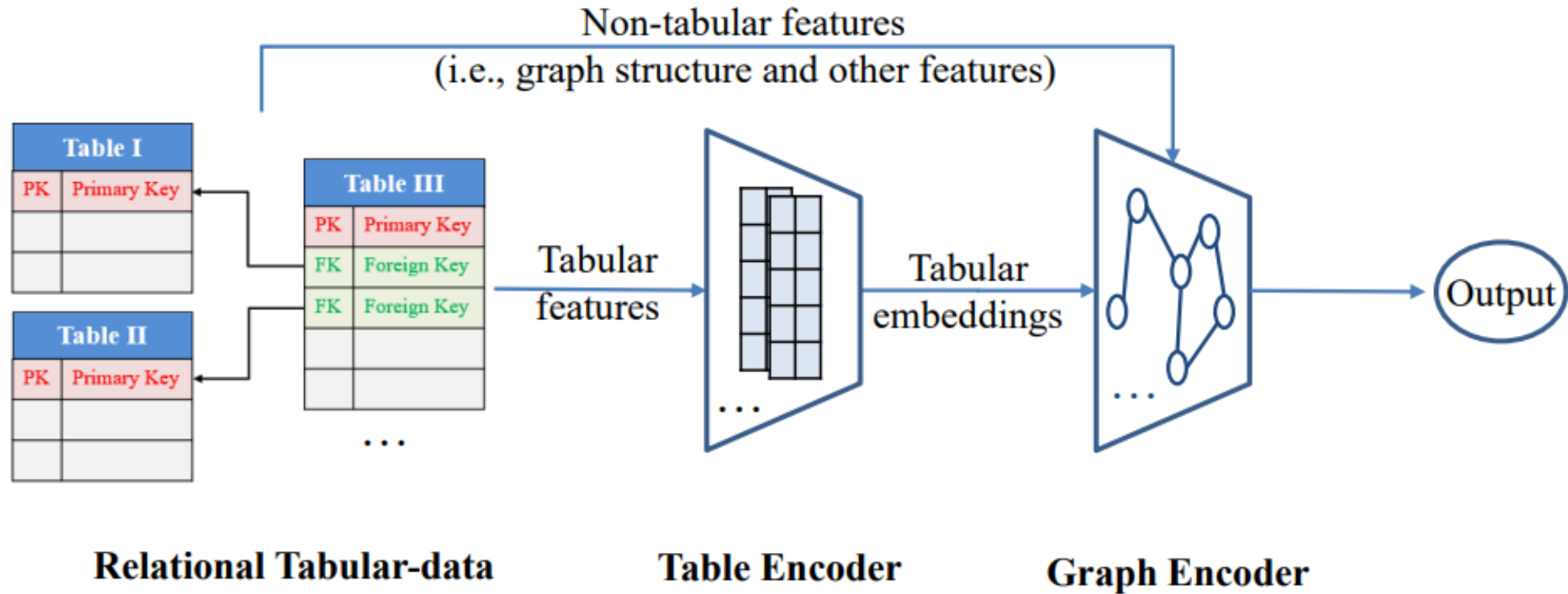


Figure 4: The architecture of BRIDGE

BRIDGE utilizes TNNs to process table data and leverages the “foreign keys” in relational tables to construct relationships between table samples, which are then analyzed using GNNs.



Pseudo-code of BRIDGE

```
1  from rllm.nn.conv.graph_conv import GCNConv
2  from rllm.nn.conv.table_conv import TabTransformerConv
3
4  graph_encoder = GraphEncoder(GCNConv, ...)
5  table_encoder = TableEncoder(TabTransformerConv, ...)
6
7  class GraphTableBridge:
8      def __init__(self, table_encoder, graph_encoder):
9          self.table_encoder = table_encoder
10         self.graph_encoder = graph_encoder
11
12     def forward(self, table_data, non_table_data, adj):
13         table_embeds = self.table_encoder(table_data)
14         node_feats = COMBINE(table_embeds, non_table_data)
15         return self.graph_encoder(node_feats, adj)
```

In practical implementations, the code lines are around 40.

Without rLLM, more than 400+ lines are needed!

Table 2: Classification accuracy.

Methods\Datasets	TML1M	TLF2K	TACM12K
Random	0.144±0.01	0.091±0.03	0.075±0.00
TabTransformer	0.347±0.02	0.137±0.08	0.142±0.01
TabNet	0.259±0.08	0.135±0.03	0.120±0.02
FT-Transformer	0.352±0.02	0.132±0.01	0.128±0.01
BRIDGE	0.428±0.02	0.454±0.01	0.309±0.02

How to try



```
# cd ./examples
```

```
# set parameters if necessary
```

```
python bridge/bridge_tml1m.py
```

```
python bridge/bridge_tlf2k.py
```

```
python bridge/bridge_tacm12k.py
```

Project page: <https://github.com/rllm-project/rllm>

relationllm.readthedocs.io/en/latest/

rLLM

Search

INTRODUCTION

[Graph Data Handle](#)

[Table Data Handle](#)

[LLM Data Handle](#)

TUTORIALS

[Design of GNNs](#)

[Design of TNNs](#)

[Design of RTLs](#)

[Design of LLM Methods](#)

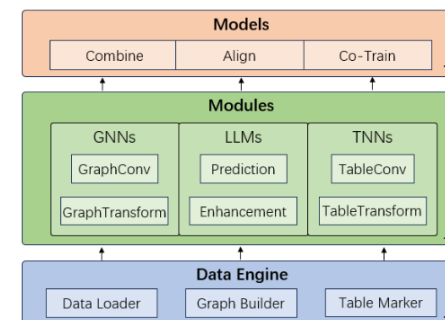
API REFERENCE

[rllm.data](#)

rLLM Documentation

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2. Facilitates novel model building in a “combine, align, and co-train” way using these modules.



The rLLM Overall Architecture

Core develop team



Zheng Wang
(Teacher)



Jianwu Zheng (PhD)



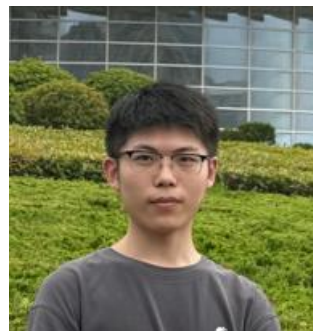
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Thanks for your time.
QA.

