

NormTab: Improving Symbolic Reasoning in LLMs Through Tabular Data Normalization



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1. Overview

We introduce **NormTab**, a framework that uses LLMs to normalize web tables into structured relational database formats, improving symbolic reasoning tasks.



2. RQs and Key Features

Research Questions:

- RQ1:** How can we leverage LLMs' textual understanding to effectively clean and normalize web tables?
- RQ2:** How can web table normalization enhance table reasoning tasks, particularly in the context of LLM based symbolic reasoning?

Key Features:

- (1) Structure normalization:** transposing tables, flattening rows and columns.
- (2) Value normalization:** removing extraneous strings, standardizing dates and numbers.
- (3)** Effective for data cleaning and transformation tasks.
- (4)** Addresses structural variance, mixed values, noise, and substring extraction in web tables

4. Experimental Setup

LLM: gpt-3.5-turbo-0125, gpt-4-turbo, gemini-1.5-flash

Datasets: (1) WikiTableQuestions (2) TabFact

Evaluation Process:

- We apply Text-to-SQL on Original tables and Normalized tables
- Compare the performance

6. NormTab Accuracy

Type	Accuracy
Columns Selection	91.0%
Transpose Detection	97.0%
Last Row Aggregation	100.0%
Split Column	87.0%
Date and Number	100.0%
N/A value	93.0%
Value Cleaning	82.0%

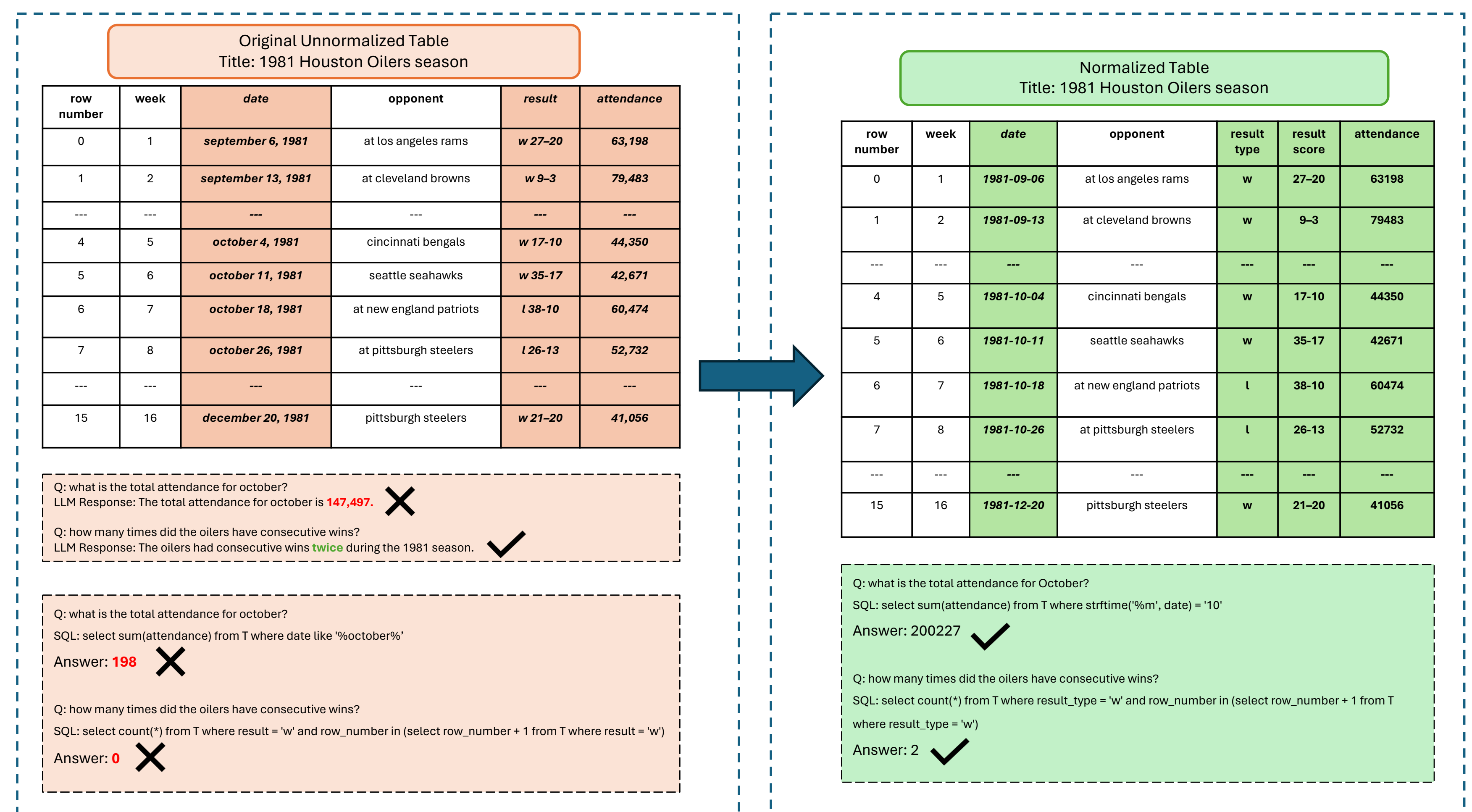
Table 4: Accuracy of **NormTab** in various types of normalization.

7. NormTab: Basic vs Targeted

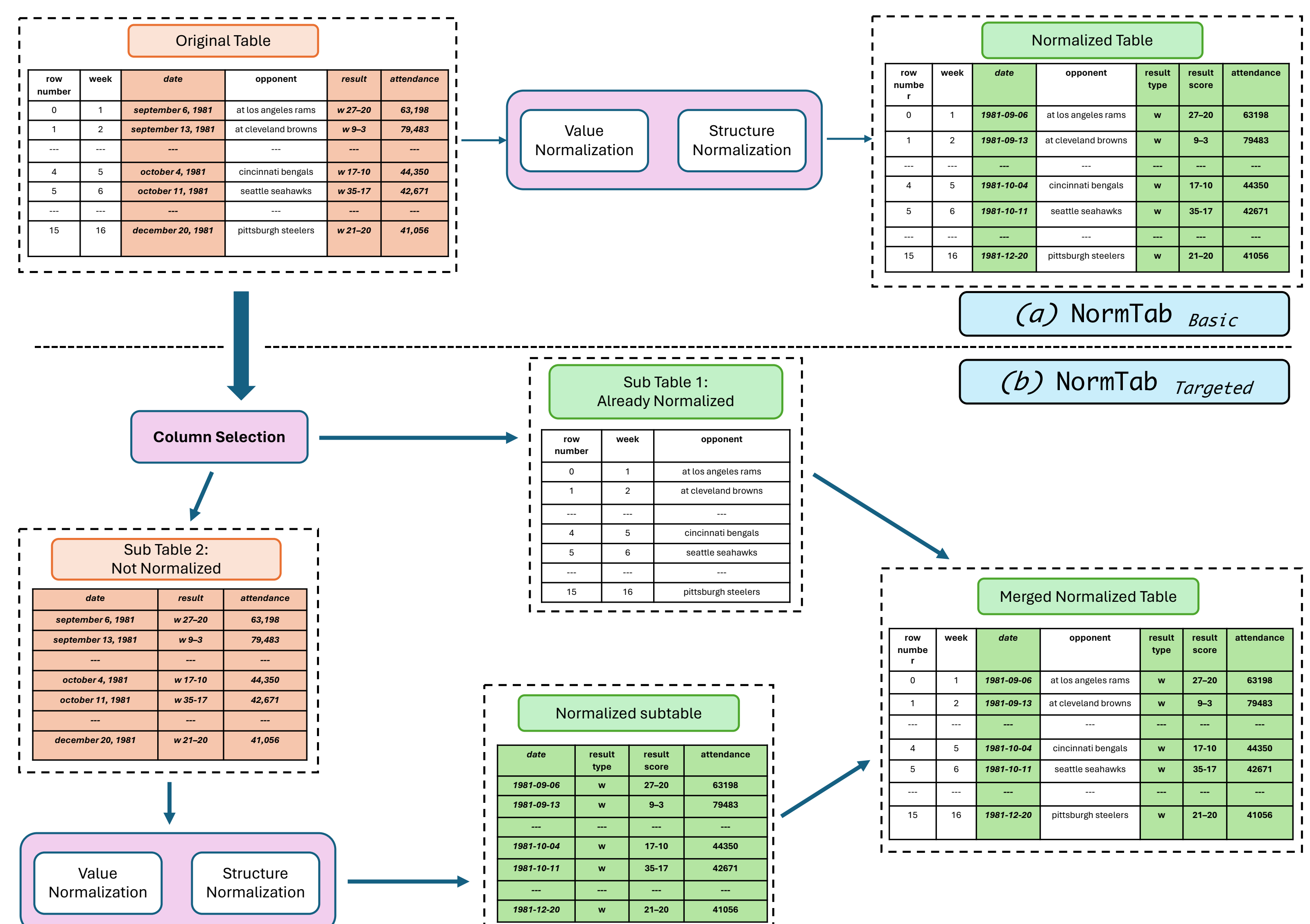
Dataset	NormTab (Basic)	NormTab (Targeted)	Reduction
WikiTQ	152.26	41.82	72.53%
TabFact	106.19	29.11	72.58%

Table 8: Table cell reduction in NormTab-Targeted compared to NormTab-Basic

3A. Motivation



3B. Methodology



5. Results

Model	Acc (%)
TableCoT (Chen, 2023)	52.40
BINDER	56.74
DATER	52.80
ReAcTable	52.40
Rethinking-Tab-Data	56.87
Chain-of-Table	59.94
Text-to-SQL (Rajkumar et al., 2022)	52.90
Text-to-SQL (gpt-3.5-turbo)	51.30
NormTab_{Basic} + SQL (ours)	60.80
NormTab_{Targeted} + SQL (ours)	61.20

Table 1: Performance comparison of NormTab on WikiTQ dataset. The results clearly demonstrate that NormTab significantly surpasses other models in accuracy when employing symbolic reasoning.

Model	Acc (%)
TableCoT-chatgpt	73.10
BINDER	79.17
DATER	78.01
Chain-of-Table	80.20
ReAcTable	73.10
Text-to-SQL (Rajkumar et al., 2022)	64.71
Text-to-SQL (gpt-3.5-turbo)	62.32
NormTab_{Basic} + SQL (ours)	67.10
NormTab_{Targeted} + SQL (ours)	68.90

Table 2: Performance comparison of NormTab on TabFact dataset with other models.

Model	Acc (WTQ)
NormTab _{Basic} (gemini-1.5-flash)	61.36
NormTab _{Targeted} (gemini-1.5-flash)	61.24
NormTab _{Basic} (gpt-4-turbo)	61.57
NormTab_{Targeted} (gpt-4-turbo)	62.28

Table 3: Performance of NormTab on WikiTQ dataset using Gemini-1.5-flash and GPT-4-turbo.

Model	Acc (WTQ)
TabSQLify	64.7
NormTab+TabSQLify	68.63

Table 5: Performance of TabSQLify integrated with NormTab on the WikiTableQuestions dataset (gpt-3.5-turbo).

8. Conclusions

Here are some **key takeaways**:

- Utilizes LLMs' text comprehension for data cleaning and normalization.
- Enhance symbolic code generation performance.
- Demonstrates significant improvement in symbolic table reasoning.