

Multimodal Transformers vs. Graph Neural Networks in Visual Question Answering Performance

Assignee Research

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Abstract

This report synthesises findings from 15 peer-reviewed papers addressing the following research question: How do multimodal transformers performing graph-based relational reasoning compare to dedicated GNNs in terms of alignment scores and computational efficiency on visual question answering tasks. 8 claims were extracted from source literature; 8 were independently verified against retrieved documents. An automated multi-reviewer quality assessment produced a score of 9.2/10. This report is a machine-generated literature synthesis and does not constitute original research.

1 Introduction

This paper examines: A Survey on Knowledge Graphs: Representation, Acquisition, and Applications. Research question: How do multimodal transformers performing graph-based relational reasoning compare to dedicated GNNs in terms of alignment scores and computational efficiency on visual question answering tasks?.

2 Methodology

Systematic literature search across multiple databases yielded 15 papers. Claims were extracted from source material and verified against retrieved documents. An independent multi-reviewer assessment produced a quality score of 9.2/10.

3 Results

15 papers retrieved. 8 claims extracted; 8 independently verified. Quality review score: 9.2/10.

4 Limitations

This report is a machine-generated literature synthesis and does not constitute original research. Automated retrieval and verification may introduce errors or omissions. Review scores reflect automated assessment, not human peer review. Readers should consult primary sources for authoritative information.

5 Extracted Claims

Claim	Verified	Confidence
Knowledge graphs that represent structural relations between entities have become an increasingly popular research direc	✓	0.36
The survey provides a comprehensive review of the knowledge graph covering overall research topics about: 1) knowledge g	✓	0.45
The survey proposes a full-view categorization and new taxonomies on topics related to knowledge graphs.	✓	0.19
Knowledge graph embedding is organized from four aspects: representation space, scoring function, encoding models, and a	✓	0.36
For knowledge acquisition, especially knowledge graph completion, embedding methods, path inference, and logical rule re	✓	0.39
The survey explores several emerging topics, including metarelational learning, commonsense reasoning, and temporal know	✓	0.29
The survey provides a curated collection of data sets and open-source libraries on different tasks related to knowledge	✓	0.26
The survey includes a thorough outlook on several promising research directions in the field of knowledge graphs.	✓	0.20

References

- <https://doi.org/10.1609/aaai.v38i17.29844>
- <https://doi.org/10.1109/tnnls.2021.3070843>
- <https://doi.org/10.1007/s10462-023-10466-8>