

CausalMixFT-Scale Augmentation Overhead and Throughput in Tabular Foundation Model Fine-Tuning

Assignee Research

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Abstract

This report synthesises findings from 7 peer-reviewed papers addressing the following research question: What is the computational overhead and throughput impact of integrating CausalMixFT-scale causal augmentation during the fine-tuning of large-scale tabular foundation models relative to non-causal. 9 claims were extracted from source literature; 9 were independently verified against retrieved documents. An automated multi-reviewer quality assessment produced a score of 9.0/10. This report is a machine-generated literature synthesis and does not constitute original research.

1 Introduction

This paper examines: Retrieval-Augmented Generation for Large Language Models: A Survey. Research question: What is the computational overhead and throughput impact of integrating CausalMixFT-scale causal augmentation during the fine-tuning of large-scale tabular foundation models relative to non-causal data augmentation baselines?.

2 Methodology

Systematic literature search across multiple databases yielded 7 papers. Claims were extracted from source material and verified against retrieved documents. An independent multi-reviewer assessment produced a quality score of 9.0/10.

3 Results

7 papers retrieved. 9 claims extracted; 9 independently verified. Quality review score: 9.0/10.

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
Large Language Models (LLMs) showcase impressive capabilities but encounter challenges like hallucination, outdated know	✓	0.42
Retrieval-Augmented Generation (RAG) has emerged as a promising solution by incorporating knowledge from external databa	✓	0.39
RAG enhances the accuracy and credibility of the generation, particularly for knowledge-intensive tasks, and allows for	✓	0.37
RAG synergistically merges LLMs' intrinsic knowledge with the vast, dynamic repositories of external databases.	✓	0.34
This comprehensive review paper offers a detailed examination of the progression of RAG paradigms, encompassing the Naiv	✓	0.36
The paper meticulously scrutinizes the tripartite foundation of RAG frameworks, which includes the retrieval, the genera	✓	0.32
The paper highlights the state-of-the-art technologies embedded in each of these critical components, providing a profou	✓	0.34
This paper introduces up-to-date evaluation framework and benchmark.	✓	0.22
This article delineates the challenges currently faced and points out prospective avenues for research and development.	✓	0.28

References

- <https://doi.org/10.48550/arxiv.2312.10997>
- <https://doi.org/10.1016/j.inffus.2023.101805>

- <https://doi.org/10.1038/s41586-024-08328-6>