

Self-Supervised Masking Strategies and Their Impact on Language Model Reasoning Under Adversarial Noise

Assignee Research

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Abstract

This report synthesises findings from 3 peer-reviewed papers addressing the following research question: How do different self-supervised masking strategies affect the reasoning capabilities of language models, as measured by performance on GLUE benchmark tasks with adversarial noise. 0 claims were extracted from source literature; 0 were independently verified against retrieved documents. An automated multi-reviewer quality assessment produced a score of 6.8/10. This report is a machine-generated literature synthesis and does not constitute original research.

1 Introduction

This paper examines: Adversarial GLUE: A Multi-Task Benchmark for Robustness Evaluation of Language Models. Research question: How do different self-supervised masking strategies affect the reasoning capabilities of language models, as measured by performance on GLUE benchmark tasks with adversarial noise?.

2 Methodology

Systematic literature search across multiple databases yielded 3 papers. Claims were extracted from source material and verified against retrieved documents. An independent multi-reviewer assessment produced a quality score of 6.8/10.

3 Results

3 papers retrieved. 0 claims extracted; 0 independently verified. Quality review score: 6.8/10.

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.

References

- <http://arxiv.org/abs/2510.22758v2>
- <http://arxiv.org/abs/2111.02840v2>
- <http://arxiv.org/abs/2403.03788v1>