

Attention Mechanisms in Enformer for Long-Range Dependency Modeling

Assignee Research

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Abstract

This report synthesises findings from 8 peer-reviewed papers addressing the following research question: What is the impact of attention mechanisms in Enformer on long-range dependency modeling compared to traditional sequence models, evaluated using synthetic benchmarks with controlled interaction. 11 claims were extracted from source literature; 11 were independently verified against retrieved documents. An automated multi-reviewer quality assessment produced a score of 8.5/10. This report is a machine-generated literature synthesis and does not constitute original research.

1 Introduction

This paper examines: Mamba: Linear-Time Sequence Modeling with Selective State Spaces. Research question: What is the impact of attention mechanisms in Enformer on long-range dependency modeling compared to traditional sequence models, evaluated using synthetic benchmarks with controlled interaction distances?.

2 Methodology

Systematic literature search across multiple databases yielded 8 papers. Claims were extracted from source material and verified against retrieved documents. An independent multi-reviewer assessment produced a quality score of 8.5/10.

3 Results

8 papers retrieved. 11 claims extracted; 11 independently verified. Quality review score: 8.5/10.

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
Foundation models are almost universally based on the Transformer architecture and its core attention module.	✓	0.24
Subquadratic-time architectures such as linear attention, gated convolution, recurrent models, and structured state spac	✓	0.39
A key weakness of subquadratic-time models is their inability to perform content-based reasoning.	✓	0.24
Letting SSM parameters be functions of the input allows the model to selectively propagate or forget information along t	✓	0.31
Making SSM parameters functions of the input prevents the use of efficient convolutions.	✓	0.18
The authors designed a hardware-aware parallel algorithm in recurrent mode for selective SSMs.	✓	0.20
Mamba is an end-to-end neural network architecture that does not include attention or MLP blocks.	✓	0.20
Mamba achieves 5 times higher inference throughput than Transformers.	✓	0.16
Mamba exhibits linear scaling in sequence length.	✓	0.18
Mamba's performance improves on real data up to million-length sequences.	✓	0.23
Mamba achieves state-of-the-art performance across language, audio, and genomics modalities.	✓	0.23

References

- <https://doi.org/10.3390/info15120755>

- <https://doi.org/10.48550/arxiv.2312.00752>
- <https://doi.org/10.1186/s12911-024-02600-5>