

# Adaptive Risk Scheduling for Language Model Generalization in Mathematical Reasoning

Assignee Research

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## Abstract

This report synthesises findings from 15 peer-reviewed papers addressing the following research question: What training strategies improve language model generalization to novel mathematical reasoning problems v7. 9 claims were extracted from source literature; 1 was independently verified against retrieved documents. An automated multi-reviewer quality assessment produced a score of 4.5/10. This report is a machine-generated literature synthesis and does not constitute original research.

## 1 Introduction

This paper examines: ARS: Adaptive Reasoning Suppression for Efficient Large Reasoning Language Models. Research question: What training strategies improve language model generalization to novel mathematical reasoning problems v7.

## 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 4.5/10.

## 3 Results

15 papers retrieved. 9 claims extracted; 1 independently verified. Quality review score: 4.5/10.

## 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 |
|---|----------|------------|
| ARS consistently achieves superior length reduction while maintaining competitive accuracy across all model scales.               | ×        | 0.11       |
| ARS operates through three core components: (1) Multi-checkpoint certainty estimation, (2) Progressive threshold adaptat          | ✓        | 0.16       |
| ARS establishes multiple checkpoints $\{c_1, c_2, \dots, c_k\}$ at regular intervals during generation.                           | ×        | 0.02       |
| At each checkpoint $c_i$ , ARS estimates model certainty through tentative answer probing.  | ×        | 0.05       |
| The heuristic difficulty estimation is used to schedule the mode of operation in ARS.   | ×        | 0.03       |
| ARS uses different policies (CoDFastPolicy, ElasticModeratePolicy, DeepReflectPolicy) based on the difficulty of the que          | ×        | 0.02       |
| ARS sets a maximum token limit of 1200 tokens per response.   | ×        | 0.03       |
| ARS aims to minimize the expected output length $E[T]$ while preserving reasoning accuracy.                                       | ×        | 0.06       |
| ARS uses specific keywords $T = \{\text{"Wait"}, \text{"But"}, \text{"Alternatively"}, \dots\}$ to identify reflection behaviors. | ×        | 0.02       |

## References

- <http://arxiv.org/abs/2510.00071v2>
- <http://arxiv.org/abs/2509.25160v1>
- <http://arxiv.org/abs/2402.05808v2>