

UNIVERSITY OF TARTU
Institute of Computer Science
Data Science Curriculum

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Prefix Tuning Applicability In Language Models

Master's Thesis (15 ECTS)

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Tartu 2026

Prefix Tuning Applicability In Language Models

Abstract:

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Keywords: LLM, fine-tuning

CERCS: P176

Prefix Tuning Applicability In Language Models

Lühikokkuvõte:

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Võtmesõnad: LLM, fine-tuning

CERCS: P176

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1. Introduction

$$h_i = \text{LM}_\phi(z_i, h_{<i})$$

2. Related Work

2.1. Context based parameter efficient fine-tuning

Prefix tuning [1], P-Tuning [2], Prompt tuning [3].

2.2. Weight based parameter efficient fine-tuning

LoRA [4].

2.3. Language models

BERT [5], Llama [6].

Few-shot learning [7].

3. Conclusion

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4. Appendices

5. License