

# Fine-Tuning Multilingual LLMs on Non-English WebFAQ Data for Zero-Shot Cross-Lingual Reasoning Transfer

Assignee Research

June 12, 2026

## Abstract

Large Language Models (LLMs) have demonstrated remarkable capabilities, yet their performance in low-resource languages (LRLs), such as Swahili, often lags due to data scarcity and underrepresentation in pre-training. A key challenge is achieving robust cross-lingual lexical alignment, crucial for tasks like translation and cross-lingual information retrieval. This paper introduces Targeted Lexical Injection (TLI), a novel and efficient fine-tuning approach. We first demonstrate that Lughu-Llama-8B-wura, a Swahili-centric LLM, exhibits strong, near-perfect lexical alignment for Swahili-English

## 1 Introduction

This paper examines: Targeted Lexical Injection: Unlocking Latent Cross-Lingual Alignment in Lughu-Llama via Early-Layer LoRA Fine-Tuning. Research question: To what extent does fine-tuning multilingual LLMs on WebFAQ's non-English QA pairs improve zero-shot cross-lingual transfer performance on reasoning benchmarks like XREX, measured by exact match accuracy?

## 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 7.0/10.

## 3 Results

15 papers retrieved. 18 claims extracted; 13 independently verified. Quality review score: 7.0/10.

## 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
Layer 0 (input embeddings) of the Lughha-Llama-8B-wura model showed an average cosine similarity of approximately 0.3153	✓	0.24
Layer 1 of the Lughha-Llama-8B-wura model showed an average cosine similarity of 0.9808 in the pilot study.	✓	0.22
Layer 2 exhibited the peak average cosine similarity of 0.99998 in the pilot study.	✓	0.19
Layer 31 showed an average cosine similarity of 0.9876 in the pilot scan.	✓	0.19
The baseline output similarity on the full evaluation set prior to TLI fine-tuning was approximately 0.32.	×	0.14
The average cosine similarity at Layer 31 for the trained set prior to TLI fine-tuning was approximately 0.3211.	✓	0.21
The average cosine similarity at Layer 31 for the control set (63 unseen pairs) prior to TLI fine-tuning was approximate	✓	0.22
The study uses Lughha-Llama-8B-wura as the base model.	×	0.15
Lughha-Llama is built upon the Llama-3 architecture.	×	0.14
The model was loaded in 4-bit precision using bitsandbytes with NF4 quantization.	✓	0.19
The compute data type used was torch.bfloat16.	×	0.12
The pilot study analyzed embeddings from Layers 0 through 31.	×	0.10
Layer 0 represents the initial input embeddings.	✓	0.19
For the evaluation phase, word embeddings were extracted from the final output layer (Layer 31).	✓	0.20
Embeddings were mean-pooled over attention-masked tokens and L2-normalized.	✓	0.21
Cosine similarity between L2-normalized Swahili and English word embeddings was used as the primary metric for lexical a	✓	0.28
The control set consisted of 63 unseen word pairs.	✓	0.16
A paired t-test was conducted to determine the statistical significance of changes in mean cosine similarity before and	✓	0.25

## References

- <http://arxiv.org/abs/2312.10793v3>
- <http://arxiv.org/abs/2506.15415v1>
- <http://arxiv.org/abs/2005.00633v1>