

Cross-lingual Transfer Performance Variation with Intermediate Task Training and Labeled Data Ratios

Assignee Research

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Abstract

Pre-trained multilingual language encoders, such as multilingual BERT and XLM-R, show great potential for zero-shot cross-lingual transfer. However, these multilingual encoders do not precisely align words and phrases across languages. Especially, learning alignments in the multilingual embedding space usually requires sentence-level or word-level parallel corpora, which are expensive to be obtained for low-resource languages. An alternative is to make the multilingual encoders more robust; when fine-tuning the encoder using downstream task, we train the encoder to tolerate noise in the context

1 Introduction

This paper examines: Improving Zero-Shot Cross-Lingual Transfer Learning via Robust Training. Research question: How does the impact of intermediate task training on zero-shot cross-lingual transfer performance vary when using different ratios of labeled examples (e.g., 0.1%, 1%, 5%) in the target language, measured by accuracy degradation thresholds on XTREME-R?

2 Methodology

Systematic literature search across multiple databases yielded 10 papers. Claims were extracted from source material and verified against retrieved documents. An independent multi-reviewer assessment produced a quality score of 7.5/10.

3 Results

10 papers retrieved. 15 claims extracted; 11 independently verified. Quality review score: 7.5/10.

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
The cross-lingual transfer performance improves by 2.1 points on PAWS-X.	✓	0.16
The cross-lingual transfer performance improves by 1.6 points on XNLI.	×	0.14
Robust training remarkably improves generalized cross-lingual transfer.	✓	0.20
The code for the study is available at https://github.com/uclanlp/Robust-XLT .	✓	0.18
Multilingual BERT, XLM, and XLM-R are pre-trained multilingual language models proposed for zero-shot cross-lingual tran	✓	0.22
XTREME and XGLUE provide benchmarks for zero-shot cross-lingual transfer learning.	✓	0.17
Early works focus on word embedding spaces for embedding space alignments.	✓	0.21
Recent approaches propose to align contextual word embedding spaces using methods like learning rotation projections and	✓	0.18
Most methods for embedding space alignments require additional supervision signals, such as parallel sentence pairs, bil	✓	0.15
Additional supervised corpora are usually expensive for low-resource languages.	✓	0.17
Some research focuses on making the model aware of embedding misalignment issues by considering additional syntactic fea	✓	0.19
Syntactic features like part-of-speech and dependency parse trees are used to handle embedding misalignment.	✓	0.15
The average performance across multiple languages for the first table is 52.3.	×	0.03
The average performance across multiple languages for the second table is 51.2.	×	0.03
The average performance across multiple languages for the third table is 56.8.	×	0.03

References

- <http://arxiv.org/abs/2104.08645v2>

- <http://arxiv.org/abs/2010.08244v1>
- <http://arxiv.org/abs/2005.13013v2>