

Ktransfer Project Overview

Exploring Cross-Domain Few-Shot
Prompting

Introduction

- Objective: Develop an effective retriever for **cross-domain** few-shot prompting.
- Background: Few-shot examples from a **hint domain** aiding a **test domain**.

Methodology

- Data Collection:
 - Randomly sampled hints from hint domains.
 - Processed to generate positive/negative samples.
- Model Training:
 - Meta-network with Transformer-based sub-networks.
 - Loss: Pairwise contrastive loss.

Experiment Setup

- Dataset: MMLU split into **hint** and **test** domains.
- Model:
 - Input: 1536 dimensions (embedding).
 - Layers: 2 Transformer layers, 8 heads.
 - Batch size: 1024; Learning rate: $1e-6$; Epochs: 10.

Results

- Cosine Similarity:
 - No significant improvement for most test domains.
- Improved Retriever:
 - Outperformed random selection in some scenarios.

Analysis & Future Work

- Analysis:
 - Current retriever performance limited in complex scenarios.
 - Need for advanced semantic models.
- Future Work:
 - Optimize model structure.
 - Explore more robust training techniques.

Conclusion

- Key Findings:
 - Pairwise-trained retriever shows potential.
 - Challenges in generalizing to complex domains.
- Outlook:
 - Further refinements needed for broader applicability.

Q&A

- Thank you for your attention! Feel free to ask questions.