Hindi-Chhattisgarhi Cross-Lingual Transfer Using RL-Guided Distillation

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1 Project Overview

This approach focuses on developing a specialized framework that efficiently transfers knowledge from Hindi (high-resource language) to Chhattisgarhi (low-resource language) using reinforcement learning and knowledge distillation techniques. We concentrate resources on optimizing core transfer learning mechanisms.

2 Key Components

- NLLB-Enhanced Transfer Learning: Utilizing the No Language Left Behind dataset as the foundation.
- Reinforcement Learning for Selective Knowledge Transfer: Using RL to dynamically determine optimal transfer parameters.
- Bidirectional Lexical Mapping: Leveraging existing Hindi-Chhattisgarhi parallel data.
- Knowledge Distillation: Implementing confidence-guided distillation techniques.

3 Project Timeline

3.1 Phase 1: Setup and Foundational Development (March 13-20, 2025)

• Environment Setup and Data Collection

- Configure development environment with required libraries and frameworks.
- Access and prepare the NLLB dataset components relevant to Hindi.
- Collect and organize existing Hindi-Chhattisgarhi parallel corpus (40,000 sentences).
- Set up evaluation metrics and benchmarks for cross-lingual transfer.

• Base Model Architecture Design

- Design and implement the base encoder-decoder architecture using NLLB Hindi components.
- Establish baseline model performance metrics.
- Create data preprocessing pipelines for both Hindi and Chhattisgarhi corpora.

• Knowledge Distillation Framework Implementation

- Develop teacher and student model architectures for knowledge distillation.
- Implement confidence-guided distillation mechanisms based on DRL-Rec principles.
- Design output-level distillation using list-wise KL divergence loss.
- Implement intermediate-level distillation using Hint loss.

3.2 Phase 2: RL Framework Development (March 21-28, 2025)

• RL Environment Design

- Define state space (model parameters, performance metrics).
- Define action space (parameter selection for transfer: freeze/fine-tune/replace).
- Implement reward function based on translation quality metrics (BLEU, METEOR).

• Exploring and Filtering Module Implementation

- Develop the exploring mechanism to identify valuable transfer candidates.
- Implement filtering strategies to select informative training instances.
- Design confidence-guided filtering based on teacher model certainty.

• Policy Network Development and Integration

- Implement RL policy network using Proximal Policy Optimization (PPO).
- Integrate policy network with the transfer learning framework.
- Establish monitoring metrics for RL agent performance.

3.3 Phase 3: Training and Optimization (March 29-April 5, 2025)

- Initial Model Training and Transfer Strategy Optimization.
- RL-Guided Distillation Process and extensive ablation studies.

3.4 Phase 4: Evaluation and Documentation (April 6-13, 2025)

- Comprehensive evaluation on Hindi-Chhattisgarhi test sets.
- Drafting research paper and final documentation.

4 Project Repository Structure

```
rich-cousin-poor-sister/
|--- README.md
|--- LICENSE
|--- requirements.txt
|--- .gitignore
|--- data/
   |--- raw/
   |--- processed/
   |--- embeddings/
   |--- parallel_corpus/
    evaluation/
|--- src/
   |--- models/
   |--- preprocessing/
   |--- training/
   |--- distillation/
   |--- reinforcement_learning/
   |--- evaluation/
    utils/
|--- experiments/
   |--- logs/
```

```
|--- checkpoints/
configs/
|--- results/
|--- figures/
|--- tables/
metrics/
|--- docs/
|--- paper/
|--- methodology/
tutorials/
tests/
|--- unit_tests/
integration_tests/
```

5 Expected Outcomes and Novel Contributions

- First RL-Guided Transfer Learning System for Hindi-Chhattisgarhi.
- Enhanced Distillation Techniques.
- NLLB Adaptation for Non-Included Languages.
- Quantitative Analysis of Hindi-Chhattisgarhi Transfer.

By focusing resources on core components, this approach has the goal of supporting the "poor sister" language through knowledge transfer from its "rich cousin" while offering a streamlined and focused implementation path.