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Beary Language Translator

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Abstract: The Beary language, a unique Indo-Aryan linguistic blend spoken by the Beary Muslim community in the southwestern region of India, faces challenges of preservation and accessibility in the digital era. Despite its rich cultural and historical significance, the language has limited digital representation and tools for translation. This project, Beary Language Translator, aims to bridge this gap by developing a system that facilitates the translation of Beary into widely spoken languages such as English and Kannada, and vice versa.

The project leverages advanced natural language processing (NLP) techniques, including machine learning algorithms and linguistic modeling, to build an accurate and user-friendly translation tool. Comprehensive data collection, including Beary vocabulary, grammar, and idiomatic expressions, underpins the system's development. The translator is designed to promote language preservation, aid linguistic research, and enhance accessibility for the Beary-speaking community and those interested in learning the language.

Index Terms - Translation Tool, Natural Language Processing (NLP), Linguistic Preservation, Machine Learning, Underrepresented Languages, Digital Accessibility, Language Digitization, Cross-Language Communication, Cultural Heritage.

I. INTRODUCTION

Language is a vital medium for preserving culture, heritage, and identity. The Beary language, primarily spoken by the Beary Muslim community in the southwestern coastal regions of India, is an integral part of their rich cultural fabric. It is an Indo-Aryan language with influences from Arabic, Kannada, and Malayalam, reflecting the historical and cultural interactions of the region. Despite its uniqueness, the Beary language faces challenges in the modern digital era due to limited representation in linguistic tools and digital platforms.

The Beary Language Translator project aims to address these challenges by developing a robust and user-friendly translation tool. This tool will enable translations between Beary and widely spoken languages such as English and Kannada, fostering accessibility and encouraging its use in various contexts. By leveraging advanced natural language processing (NLP) techniques and machine learning models, this project intends to preserve and promote the language, ensuring it remains relevant in a rapidly evolving digital landscape. This initiative is not only a step towards preserving an endangered language but also an effort to bridge linguistic gaps, promote cultural inclusivity, and contribute to the broader goals of language digitization and preservation. Through this project, the Beary language is poised to gain wider recognition and accessibility in the global context.

II. OBJECTIVES

- [1] **Develop a Translation Tool:** Create a reliable and efficient translator to convert text between Beary and other widely spoken languages, such as English and Kannada.
- [2] **Preserve the Beary Language:** Digitize Beary vocabulary, grammar, and idiomatic expressions to aid in the preservation and promotion of the language.
- [3] **Enhance Accessibility:** Facilitate easier understanding and communication for both native Beary speakers and individuals interested in learning the language.
- [4] **Promote Cultural Heritage:** Highlight and celebrate the linguistic and cultural significance of the Beary language by making it more accessible globally.
- [5] **Leverage Modern Technologies:** Utilize advanced natural language processing (NLP) and machine learning techniques to build an accurate and scalable translation system.
- [6] **Contribute to Linguistic Research:** Provide valuable resources and insights for linguists and researchers working on underrepresented languages.
- [7] **Encourage Community Engagement:** Involve the Beary-speaking community in data collection, testing, and feedback to ensure the tool aligns with cultural and linguistic nuances.
- [8] **Foster Inclusivity:** Bridge linguistic divides by enabling seamless communication across languages, thus promoting inclusivity and mutual understanding.

III. Review Of Related works

1. "Bridging the Gap: Towards Linguistic Resource Development for Lambani"

Proposed System: This study outlines a systematic approach to developing linguistic tools for the Lambani language, encompassing data collection, preprocessing, translation to a contact language, manual part-of-speech tagging, and morphological analysis.

Research Gap: The research addresses the scarcity of linguistic resources for Lambani, a low-resource language, by creating foundational tools to facilitate further linguistic analysis and preservation efforts.

2. "LLM for Everyone: Representing the Underrepresented in Large Language Models"

Proposed System: This thesis evaluates the capabilities of large language models (LLMs) in underrepresented languages, identifying challenges in multilingual and multicultural generalization.

Research Gap: The study highlights the limitations of LLMs in effectively processing underrepresented languages, emphasizing the need for more inclusive NLP research and development.

3. "Diversity and Language Technology: How Language Modeling Bias Causes Epistemic Injustice"

Proposed System: The paper examines biases in AI-based language technologies, such as large language models and machine translation systems, which predominantly support a small fraction of the world's languages.

Research Gap: It underscores the digital language divide and the resulting epistemic injustice, calling for efforts to extend language technology support to a broader array of languages.

4. "Overcoming Language Barriers in Academia: Machine Translation Tools and Multilingual Dissemination"

Proposed System: This article discusses the role of machine translation tools in bridging language gaps in scientific communication, promoting multilingual dissemination of research.

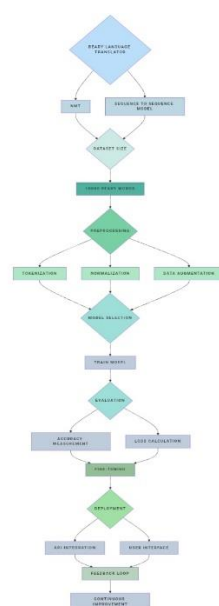
Research Gap: The work identifies challenges in achieving accurate translations that preserve contextual and cultural nuances, particularly for underrepresented languages.

5. "Multilingual Machine Translation with Large Language Models: Empirical Results and Observations"

Proposed System: The study evaluates the performance of large language models, such as GPT-4, in multilingual machine translation tasks, comparing them with supervised baselines like NLLB.

Research Gap: It reveals that while LLMs show promise, there remains a significant performance gap, especially for low-resource languages, indicating the need for further research and development.

IV. PROPOSED MODEL ARCHITECTURE



1. Start

The project begins with defining its primary goal: creating a translator for the Beary language to bridge linguistic gaps. This involves outlining the scope, identifying objectives, and planning the required resources and methodologies.

2. Approach

Selection

Neural Machine Translation (NMT) and Sequence-to-Sequence (Seq2Seq) models are evaluated as potential approaches for the translator. The choice depends on factors like the complexity of the Beary language and the expected accuracy of translation.

3. Dataset

Preparation

A dataset of 10,000 Beary words is collected to form the basis for training the translation model. This dataset incorporates vocabulary, grammar structures, and common phrases to capture the language's diversity.

4. Preprocessing

Preprocessing includes tokenization to divide text into manageable units and normalization to standardize the dataset. Data augmentation is applied to expand the dataset and improve model training by introducing variations in input text.

5. Model

Selection

The most suitable model is selected to effectively handle the unique features of the Beary language. The decision considers factors like computational efficiency, compatibility with the dataset, and expected performance.

6. Model

Training

The chosen model is trained on the prepared dataset, enabling it to learn the mapping between Beary and the target languages. Training involves optimizing the model's parameters to enhance its accuracy in translation.

7. Evaluation

The model's performance is evaluated using metrics like accuracy and loss to assess its translation quality. Any identified shortcomings are addressed through further fine-tuning and optimization.

8. Deployment

The final model is deployed, integrating APIs for developers and a user-friendly interface for end-users. A feedback loop is established to continuously improve the system based on user input and performance reviews.

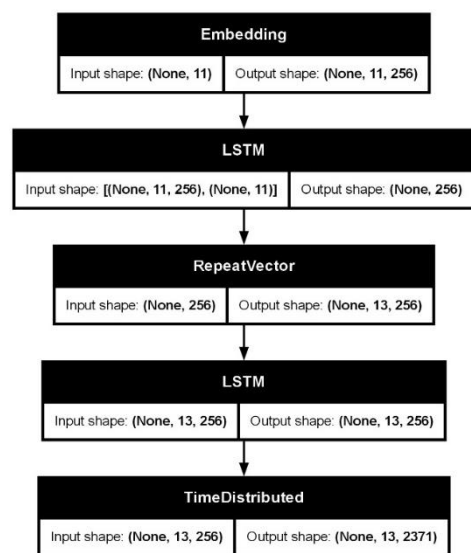
VI.IMPLEMENTATION

The implementation begins with dataset collection, where 10,000 Beary words, phrases, and their translations are gathered from various sources, including community input and linguistic texts. The data is annotated and validated to ensure accuracy and represent diverse linguistic structures. Preprocessing is then performed, involving tokenization to divide text into manageable units, normalization to standardize the dataset, and data augmentation to enhance the dataset size and variety.

The next step is model selection, where either Neural Machine Translation (NMT) or Sequence-to-Sequence models are chosen based on their compatibility with low-resource languages. Transformer-based architectures like BERT or GPT may also be used for improved translation accuracy. The selected model is trained on the preprocessed dataset, optimizing parameters through iterative learning to achieve accurate translation between Beary and the target languages.

After training, the model undergoes evaluation to assess its performance using metrics like accuracy and loss. Any issues are resolved through fine-tuning to improve the model's output. Finally, the model is deployed with API integration for developers and a user-friendly interface for end-users. A feedback loop is established to incorporate user suggestions, ensuring continuous improvement of the system.

VI.RESULTS



```

# test on some test sequences
print("test")
evaluate_model(model, eng_tokenizer, testX, test)

...
train
src=[restaurant tea cup cost encha], target=[how much is a cup of tea at the restaurant], predicted=[how much does a bottle of cost the the the]
src=[train station taxi price encha], target=[how much is a taxi to the train station], predicted=[how much is a taxi to the station station]
src=[bottle water thagolbe], target=[can i get a bottle of water], predicted=[can i get a bottle of water]
src=[groceries cash thagolbe], target=[can i pay for the groceries with cash], predicted=[can i pay for my my cash]
src=[nearby good vegetarian restaurant thakolbe encha ithhiri], target=[where can i find a good vegetarian restaurant nearby], predicted=[where can i find a good vegetarian restaurant]
src=[aysira], target=[wealth], predicted=[wealth]
src=[bahu thanda aayi baithu], target=[its a bit chilly outside], predicted=[its a bit chilly outside]
src=[enke souvenir madke beku], target=[i want to buy some souvenirs], predicted=[i need to buy some snacks]
src=[nearest gas bunk yethe undu], target=[where is the nearest gas station], predicted=[where is the nearest gas station]
src=[nearest vegan ice cream shop encha ithhiri], target=[where is the nearest vegan ice cream shop], predicted=[where is the nearest vegan ice cream shop]
BLEU-1: 0.771389
BLEU-2: 0.691139
BLEU-3: 0.654764
BLEU-4: 0.571661
test
src=[har the drink thagolbe], target=[can i buy a drink from the bar], predicted=[can i buy a drink drink the bar]
src=[ee itenge encha price undu], target=[how much for this item], predicted=[how much this this]
src=[museum opening time encha], target=[what time does the museum open], predicted=[what time does the market close]
src=[nearest shopping mallge taxi thagolbe], target=[can i get a taxi to the nearest shopping mall], predicted=[can i get a taxi to the nearest shopping mall]
src=[adhre neeru sigatte], target=[is it possible to walk there], predicted=[is it it to the there]
src=[adka], target=[near], predicted=[that]
src=[naan phone charge maddhene], target=[i need to charge my phone], predicted=[i need to my my phone]
src=[nearby bus stop thakolbe encha ithhiri], target=[where can i find a bus stop nearby], predicted=[where can i find a bus stop here]
src=[airport train ticket cost encha], target=[how much does a train ticket to the airport cost], predicted=[how much does a ticket ticket to the airport]
...
BLEU-1: 0.694741
BLEU-2: 0.689053
BLEU-3: 0.575159
BLEU-4: 0.491187
Output is truncated. View on a scrollable element or open in a text editor. Adjust cell output settings...
  
```

VII.APPLICATIONS

1. Language Preservation:

The translator helps in preserving the Beary language by digitizing its vocabulary, grammar, and expressions, making it more accessible to future generations.

2. Cultural Promotion:

It serves as a tool for promoting the Beary language and its cultural heritage, increasing awareness among non-native speakers and fostering linguistic diversity.

3. Educational Tool:

The translator can be used in educational settings to teach Beary to students and individuals interested in learning the language.

4. Cross-Language Communication:

It enables seamless communication between Beary speakers and speakers of other languages, such as English or Kannada, improving interaction in multicultural environments.

5. Supporting Linguistic Research:

Researchers can use the translation tool to study the Beary language in comparison with other languages, aiding in linguistic analysis and language development.

VIII.CONCLUSION

The Beary Language Translator project is an essential step toward preserving and promoting the Beary language in a digital world. By leveraging advanced machine learning techniques such as Neural Machine Translation (NMT) and Sequence-to-Sequence models, the translator enables accurate and efficient communication between Beary and widely spoken languages like English and Kannada. This not only aids in the language's preservation but also fosters greater cultural understanding and inclusivity.

The project addresses key challenges associated with underrepresented languages, such as the lack of digital resources and linguistic tools, ensuring that the Beary language can be integrated into modern digital platforms. Furthermore, it supports educational initiatives, linguistic research, and practical applications in government, tourism, and cross-cultural communication. With ongoing development and a feedback-driven approach, the Beary Language Translator holds the potential to contribute significantly to the revitalization and global recognition of the Beary language, ensuring its relevance for future generations.

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