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ABSTRACT

Background: Objective: Method: Results:

CCS CONCEPTS

• Social and professional topics \rightarrow Computing education; • Computing methodologies \rightarrow Artificial intelligence.

ACM Reference Format:

1 INTRODUCTION

Bhattacharyya et al. [1] characterized the challenges of NLP in Indic languages as,

- (1) **Scale and Diversity** Indic languages encompass a vast array of languages and dialects, belonging to multiple linguistic families and written in numerous distinct scripts.
- (2) Longer Utterances Sentences in Indic languages are often longer and more complex than in English, complicating tasks like parsing and speech recognition.
- (3) **Code Mixing** The frequent mixing of multiple languages in a single sentence or conversation is a common challenge in computational linguistics for the region.
- (4) **Resource Scarcity** Many Indic languages lack sufficient annotated datasets for building robust NLP and speech tools.
- (5) Absence of basic speech and NLP tools Foundational tools like morphology analyzers and speech recognition systems are either unavailable or lack accuracy for most Indic languages.
- (6) **Absence of linguistics knowledge** A limited understanding of the linguistic structure of many regional languages hinders the development of computational models.

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- (7) **Script complexity and non-standard input mechanisms**The diversity of scripts and their associated vowel and consonant combinations make input systems slower and less
- sonant combinations make input systems slower and less intuitive.(8) Non-standard transliteration Roman transliteration of
- [8] Non-standard transliteration Roman transliteration of Indic languages lacks standardization, leading to multiple ways of representing the same word.
- (9) Non-standard storage Variations in how characters are encoded and stored pose issues in data sharing and tool interoperability.
- (10) Man-made Problems Government-imposed standard keyboards and inadequate funding often stifle innovation and efficiency in linguistic computing.
- (11) Some challenging language phenomena Features like free word order, agglutination, and context-dependent pronunciation introduce additional computational hurdles.

In Jordan et al. [2], they found that amoung the languages under their consideration, English, Spanish, Vietnamese, and Tamil, the worst performing language was Tamil, generating incorrect solutions and non-sensiable translations. Though the authors did not systematically investigate the the nature of the poor performance or methods of improvement, they did note several possible reasons:

- (1)
- (2) Compared to English and Spanish it was likely a lower resourced language in the training data.

The first of these two possibilities is, perhaps, the most interesting. In Tamil, there are two core dialects which differ significantly: literary Tamil (sen-Tamil) and colloquial (kodun-Tamil) []. Additionally, given Tamil is spoken not just in and around Tamil Nadu, but also in Sri Lanka, Malaysia, and Singapore, there are a number of regional dialects [].

2 BACKGROUND

2.1 Indic LLMs and Datasets

Ramesh et al. [3] introduced Samanantar, a large-scale multilingual dataset for Indic languages containing sentence pairs in 11 languages.

- 3 METHODS
- 4 RESULTS
- 5 DISCUSSION
- **6 LIMITATIONS**
- 7 CONCLUSION

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