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ABSTRACT

There is a critical need for translation checking to localize important content across all the industries. This poster presents question-answering techniques to check the comprehensibility of a text translation. It will save several manual checking hours and will help localize text content faster and more accurately.

INTRODUCTION

- ❖ Translation checking is a significant limiting factor on the pace of localizing important content.
- ❖ There are very few available methods to test the quality of translation, which results in local language speakers lacking timely, important information in the languages they value most.
- ❖ This project will demonstrate that **question answering techniques can be used to automatically check the comprehensibility of a text translation**. In doing this, the project addresses the translation checking bottleneck.

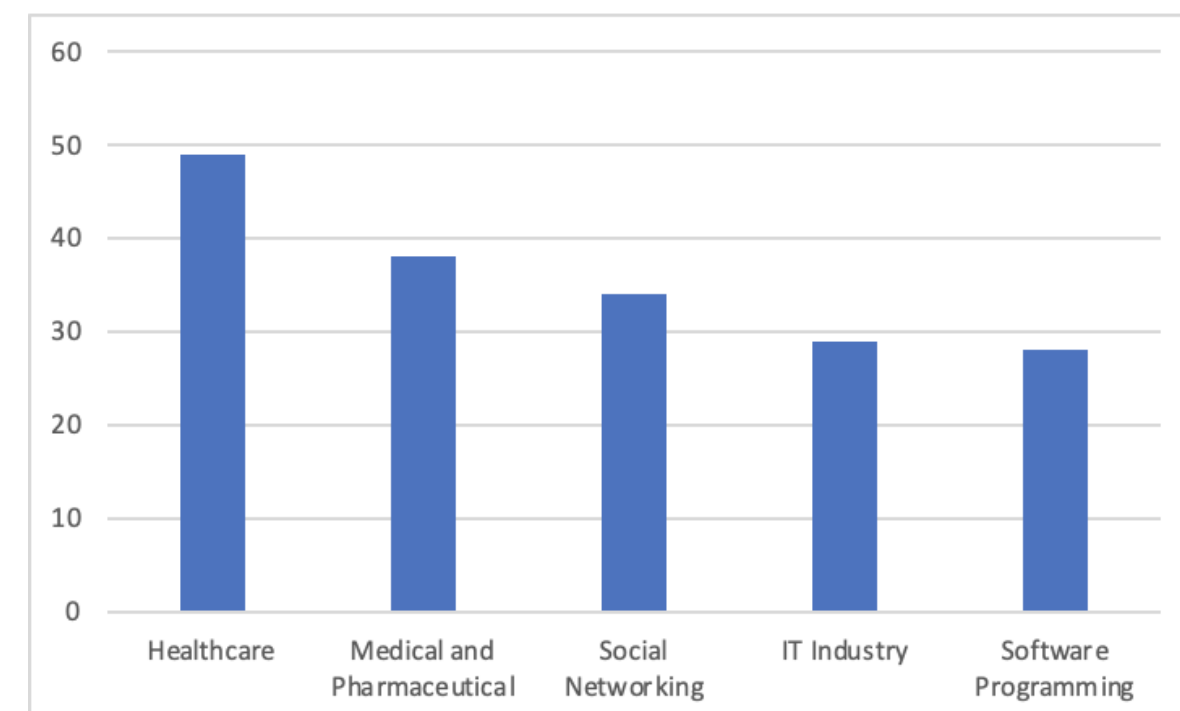
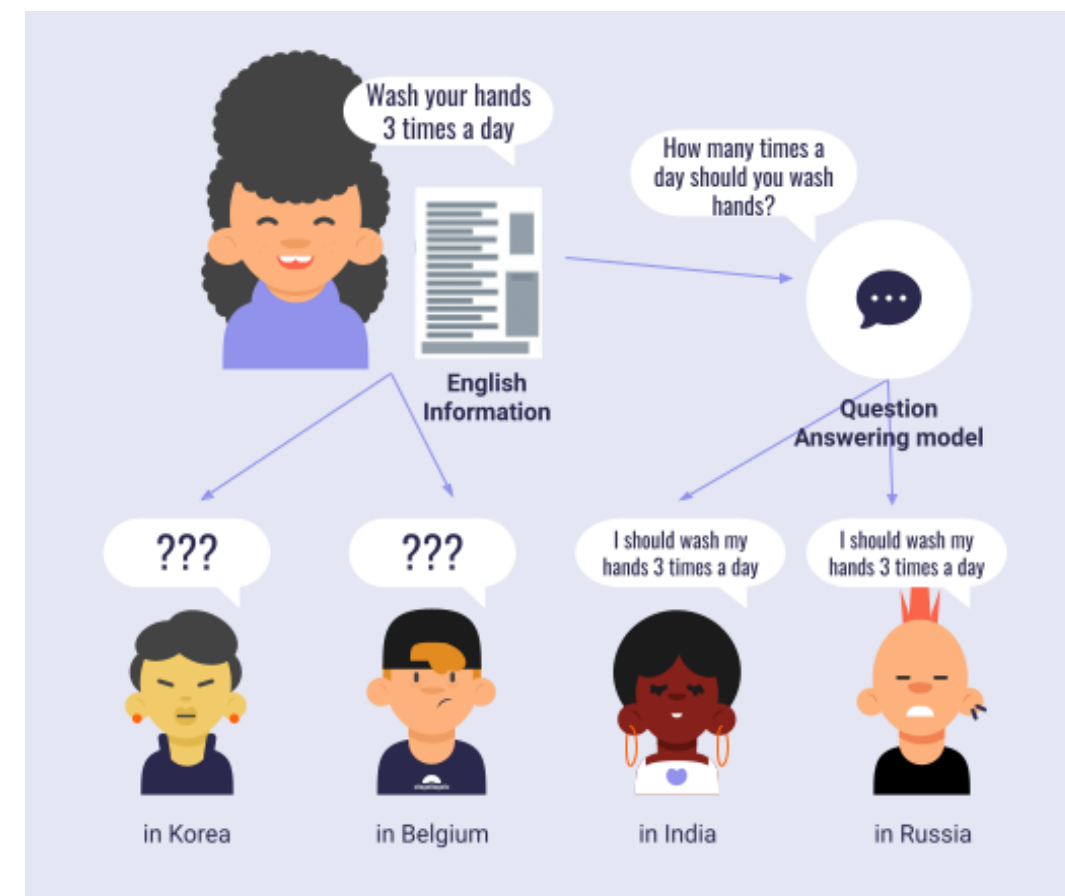


Fig. 1. Demand for translation by domain



RESEARCH QUESTIONS

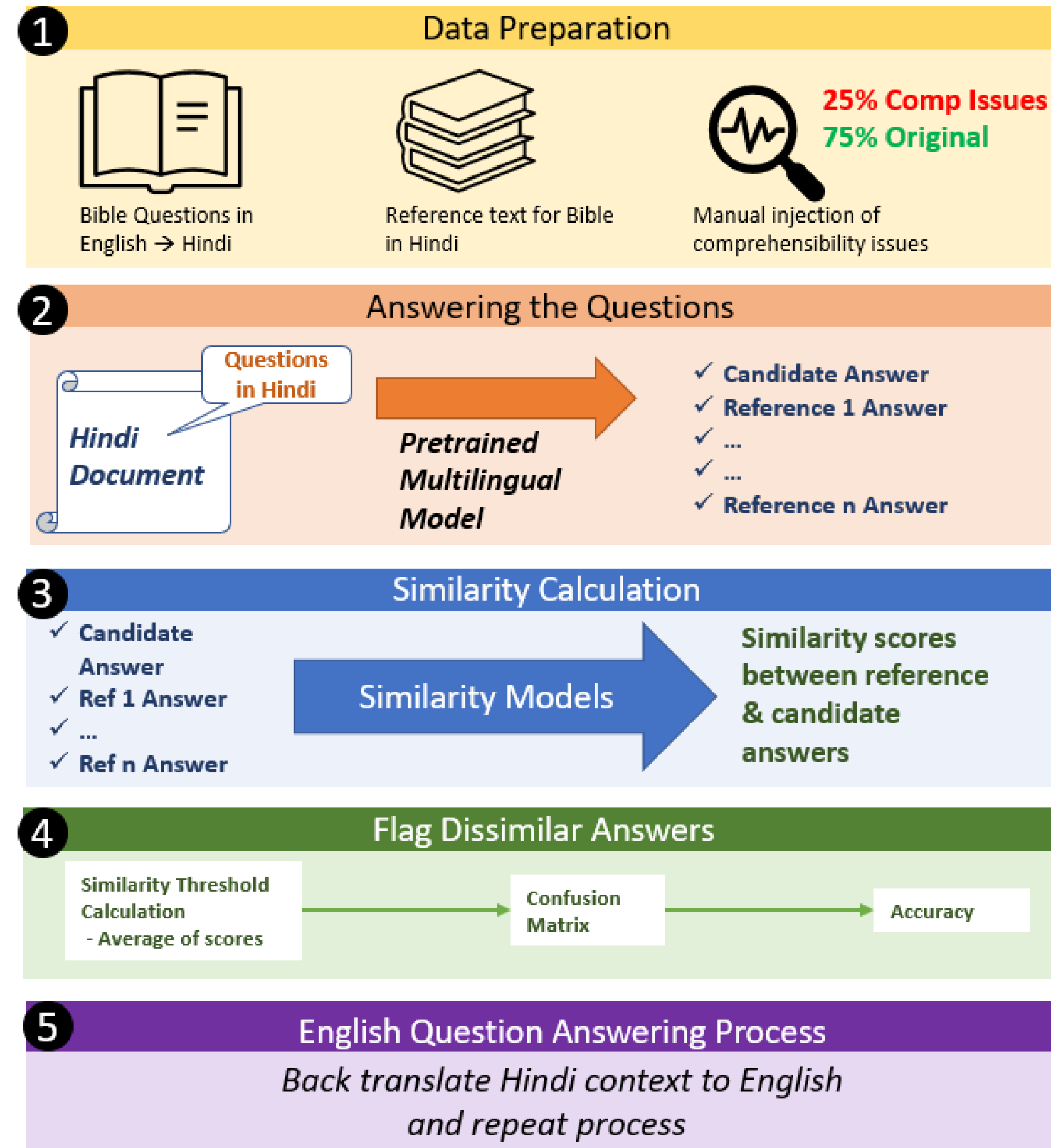
Our research focuses on answering the following questions:

- ❖ Can we create a Hindi data set for benchmarking the flagging of comprehensibility issues?
- ❖ Can we evaluate the comprehensibility of Hindi text using a pre-trained Hindi question answering model?
- ❖ Can we evaluate the comprehensibility of Hindi text using a pre-trained English question answering model and a Hindi-to-English machine generated back translation?

LITERATURE REVIEW

Study	Question-Answering	Machine translation	Similarity Calculation
Amazon Alexa Models	✓	✓	
A Qualitative Comparison of CoQA, SQuAD 2.0 and QuAC	✓	✓	
FQuAD: French Question Answering Dataset	✓		✓

METHODOLOGY



Pretrained Multilingual Question Answering Models Used –

- ❖ Bert-multi-cased-finetuned-xquadv1
- ❖ Monsoon-nlp/hindi-bert
- ❖ Monsoon-nlp/hindi-tpu-electra

Similarity Calculation Models Used –

- ❖ Language-Agnostic BERT Sentence Embedding (LaBSE)
- ❖ Bilingual Evaluation Understudy (BLEU)
- ❖ Language –Agnostic Sentence Representation (LASER)

Process of calculating Similarity Threshold–

- ❖ Take average of all similarity scores
- ❖ Determine quartiles and choose boundary
- ❖ Flag context whose score falls below threshold
- ❖ Create confusion matrix to judge accuracy

STATISTICAL RESULTS

Text_Similarity_Method	TN	FP	FN	TP	Precision	Recall	Accuracy
BLEU (AND)	169	59	31	43	42%	58%	70%
BLEU (Avg)	147	81	16	58	42%	78%	68%
LaBSE (OR)	166	62	15	59	49%	80%	75%
LaBSE (Avg)	206	22	17	57	72%	77%	87%
Eng_LaBSE (OR)	113	115	4	70	38%	95%	61%
Eng_LaBSE (Avg)	185	43	12	62	59%	84%	82%

Result Summary:

- ❖ **Hindi Question answering model** for Hindi text: BERT (base-multilingual-cased) fine-tuned for multilingual Q&A
- ❖ **English question answering model** for Back-translated English text: DistilBERT base uncased distilled SQuAD
- ❖ **Similarity Measure:** LaBSE word embedding
- ❖ **Flagging Method:** Averaging of similarity scores
- ❖ **Highest Recall rate:** using DistilBERT base uncased distilled SQuAD model on Back translated English text using average LaBSE similarity

BUSINESS IMPACTS

- ❖ Going forward, potential savings to be realized by SIL will be **\$3000 per project (50+ man hours)**.
- ❖ **A set of power tools for translation consultants**, which will allow them to do their checking work more thoroughly and more consistently
- ❖ **An early warning system for translators**, which will alert them to obvious problems so they can address them earlier in the translation process
- ❖ **An equalizer for administrators and strategists**, which will allow them to compare and evaluate methodologies and products (e.g., machine translation systems) on an equal footing

CONCLUSION

- ❖ A Hindi question answering model is a valid method to determine the comprehensibility of a translation with 85% accuracy
- ❖ A question-answering model in English on back translated data from any target language can be used to evaluate the comprehensibility of a text with 90% accuracy.
- ❖ This method can now be used by several companies in order to save several manual checking hours and will help localize text content faster and more accurately.

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