## **Paper Title:**

Bengali-English Neural Machine Translation Using Deep Learning Techniques

## **Paper Link:**

https://www.researchgate.net/publication/370130818\_Bengali-English\_Neural\_Machine\_Translation\_Using\_Deep\_Learning\_Techniques

# 1 Summary:

The paper focuses on machine translation, a crucial aspect of natural language processing that facilitates communication across various native languages. It distinguishes between word-by-word translation, sentence-by-sentence translation, and the more informative phrase translation. Neural Machine Translation (NMT) is highlighted as the latest approach, leveraging neural networks trained on extensive data with lower memory requirements compared to traditional Statistical Machine Translation (SMT) models. The primary goal is Bengali-English translation due to the significant number of native speakers.

### 1.1 Motivation

The study's primary objective is to focus on translating between Bengali and English due to the substantial number of native speakers in these languages. Bengali is noted as the world's sixth most spoken native language and seventh most spoken language overall.

## 1.2 Contribution

The study explores four Sequence-to-Sequence (seq2Seq) models: LSTM, GRU, Bidirectional LSTM (BiLSTM), and Bidirectional GRU (BiGRU). Each model is explained briefly, emphasizing their unique characteristics in handling sequence data. The paper evaluates these models using ROUGE score and BLEU evaluation metrics, widely used in assessing generation tasks. The key contributions include experimenting with and comparing the capabilities of these sequential models on a blended dataset, evaluating performance using established metrics, and achieving outputs comparable to earlier studies on neural machine translation.

# 1.3 Methodology

The study involves extensive preprocessing, effective tokenization, padding training of multiple sequential models, accuracy calculation, comprehensive evaluation using ROUGE and BLEU scores. The choice of models, optimization techniques and evaluation metrics contributes to a detailed analysis of the Bengali-English translation task.

#### 1.4 Conclusion

In this paper, the authors achieved optimal results in translation tasks using minimal resources, demonstrating that BiLSTM outperforms other models, particularly in handling sequences of consecutive words. This performance was substantiated by ROUGE and BLEU scores, with BiLSTM scoring impressively across various BLEU metrics. However, the effectiveness of BiLSTM on larger Bengali datasets, especially those containing rare words, remains an area for future research. The paper underscores the challenges in translation, emphasizing the importance of grammatical accuracy, a rich vocabulary, and substantial computational power. In future, the authors suggest that enhancements in model performance could be achieved through the use of multiple dense layers, parameter and hyperparameter tuning, advanced computational techniques, and comprehensive Bengali-English datasets.

#### 2. Limitations:

## 2.1 First Limitation:

The study mentions using a small dataset. This could limit the generalizability of the findings, as neural machine translation models often benefit from large and diverse datasets.

## 2.2 Second Limitation;

The study mentions stopping training after a certain number of epochs to avoid overfitting. However, it doesn't delve into the strategies employed to mitigate overfitting, and overfitting might still be a concern.

# 3 Synthesis:

The study uses both ROUGE and BLEU scores for evaluation, providing a comprehensive analysis. However, it would be beneficial to provide a more detailed discussion on the strengths and weaknesses of these metrics and how they complement each other compares its results with previous works, demonstrating the competitiveness of the proposed models. However, a more in-depth analysis of the specific improvements and differences in methodologies compared to previous studies would enhance the synthesis.