

INDRAPRASTHA INSTITUTE of INFORMATION TECHNOLOGY **DELHI**

Text Summarization for Hindi Language

Unlocking Clarity, One Summary at a Time

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GitHub Repo

Introduction

Text summarization in Hindi is challenging due to its rich morphological structure & diverse linguistic nuances

Hindi Text Example:

"भारतीय राजनीति में उठापटक एक आम बात है। हाल ही में, विभिन्न राज्यों में चुनावी माहौल ने जनता की राय और सरकार की नीतियों के बीच गहरे मतभेदों को उजागर किया है। इन चुनावों के परिणाम ने न केवल राजनीतिक दलों की रणनीति पर प्रभाव डाला है, बल्कि आम आदमी की दैनिक जीवन और अर्थव्यवस्था पर भी गहरा असर पड़ा है। आर्थिक मंदी, नौकरी की कमी, और सामाजिक अशांति के बीच, यह महत्वपूर्ण है कि सरकार नई नीतियाँ लागू करे जो जनता के हित में हों।"

Objective



- To develop NLP models that accurately generate summaries of Hindi texts
- To enhance readability and comprehension for different use-cases such as real-time news aggregation

Methodology

Tokenization

 To convert text into a format suitable for model input, ensuring efficient handling of Hindi text

Model Approach

 Fine-tuned IndicBART and MT5. & a custom Transformer model to accommodate the syntactic and semantic intricacies of Hindi.

Model Description

Encoder

 Layers incorporating self-attention and feedforward networks, complemented by residual connections and layer normalization for stability

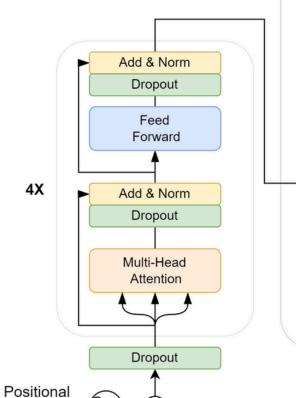
Decoder

 Layers that have dual multi-head attention mechanisms & a feed-forward network with residual connections and layer normalization

Encoding

Pre-trained Models

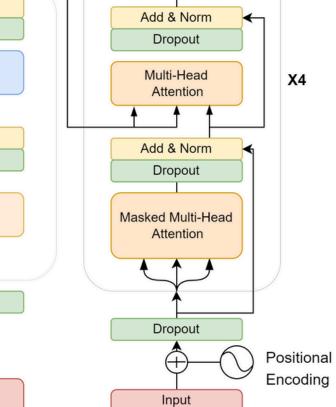
 MT5 is a multilingual textto-text model with layers and 12 attention heads



Input

Embedding

Inputs



Embedding

Target

Output

SoftMax

Linear

Add & Norm

Dropout

Feed

Forward

- IndicBART is a multilingual seq2seq model, features 6 encoderdecoder layers, 16 attention heads totaling 244M parameters
- 12 encoder-decoder each

Refrences

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Results/Findings

Model	BERTScore Precision	BERTScore Recall	BERTScore F1
IndicBART	0.789	0.852	0.819
Transformer Model	0.612	0.542	0.737
MT5 Model	0.783	0.765	0.773

QUALITATIVE ANALYSIS: T MODEL

- Gen: बहुजन पार्टी की राष्ट्रीय पार्टी की अध्यक्ष मायावती ने सत्ताईस सितंबर को किसान संगठनों द्वारा बंद बंद का समर्थन किया है
- Ref: बहजन समाज पार्टी की अध्यक्ष मायावती ने कल सत्ताईस सितंबर को किसान संगठनों द्वारा बुलाए गए भारत बंद का समर्थन

Conclusion

This project highlights IndicBART's superior performance in summarizing Hindi texts, outperforming both the Transformer and MT5 models. The results affirm the strength of fine-tuned, pre-trained models in tackling specific linguistic challenges in NLP

FUTURE WORK

- Improving the model's handling of Hindi's linguistic
- Expanding the dataset to encompass a broader spectrum of text genres for a more comprehensive performance evaluation