

# Project Presentation

CS435/635 Deep Learning



## ENGLISH TO HINDI MACHINE TRANSLATION USING TRANSFORMER

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# *Introduction*



Machine Translation (MT) is the automated process of translating text.



English and Hindi are widely spoken languages with distinct linguistic characteristics.



It provides us the opportunity to explore and implement the Transformer architecture.



Transformer model shows promising results compared to traditional statistical machine translation and other neural machine translation models.<sup>3</sup>

# ***Motivation***

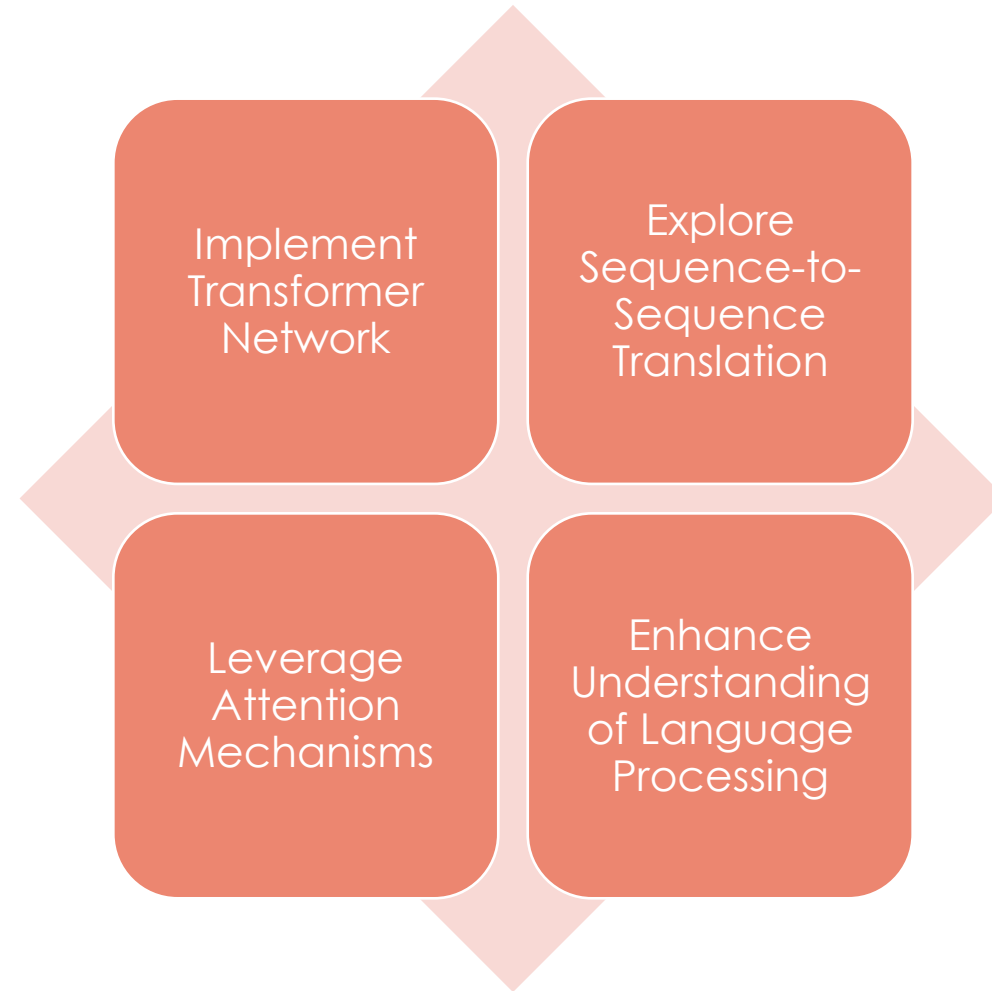
Hands-On Learning Opportunity.

Transformer's Adaptability to Sequences.

Understanding Attention Mechanisms.

Challenges of Cross-Linguistic Translation.

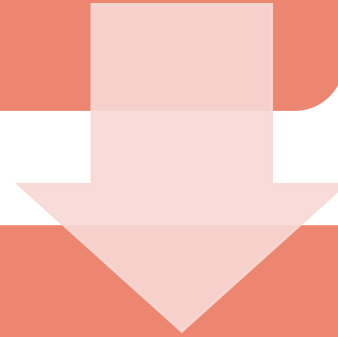
# *Objective*



# ***Methodology***

## **Data Preparation**

- Used Positional Encoding
- Tokenizer



## **Model Architecture**

- Encoder
- Decoder

# ***Methodology***

## **Training**

- Split dataset into 70% Training and 30% for testing



## **Evaluation**

- Observe Training and Testing loss



## **Optimization and Fine tuning**

- Run multiple times by changing the hyperparameter

# ***Dataset***

Used a dataset from Kaggle<sup>2</sup>.

- English to Hindi Text

The dataset contains more than 130000 records.

- Used only 100000 records.

The dataset is prepared using the corpus and scrapping the web.

The shape of dataset is:

- 130476 Row
- 2 Column



# ***Transformer Components***



Embedding  
Layer



Encoder and  
Decoder Stacks



Attention  
Mechanism



Positional  
Encoding



Feed-Forward  
Neural Networks

# *Transformer Architecture*

## Main Component

- Self-Attention Mechanism
- Multi-Head Attention
- Encoder and Decoder Stacks

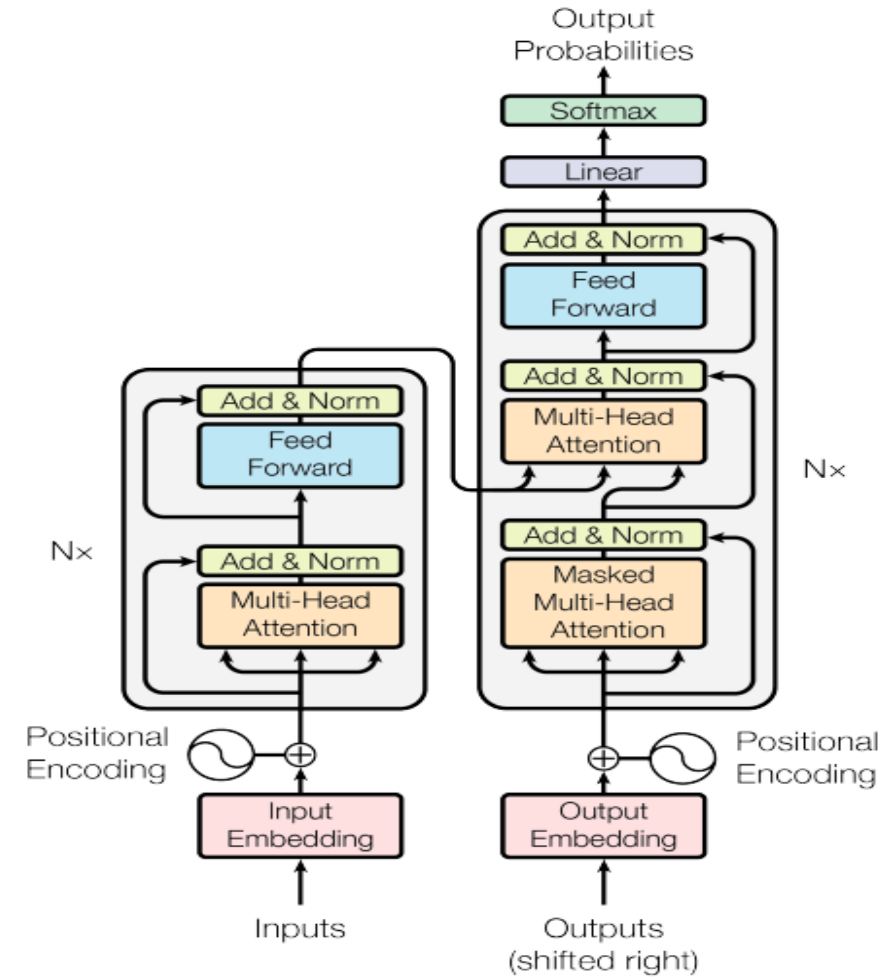


Figure 1. Attention Architecture<sup>1</sup>

# *Transformer Architecture (contd.)*

## Main Component

- Positional Encoding
- Feed-Forward Neural Networks

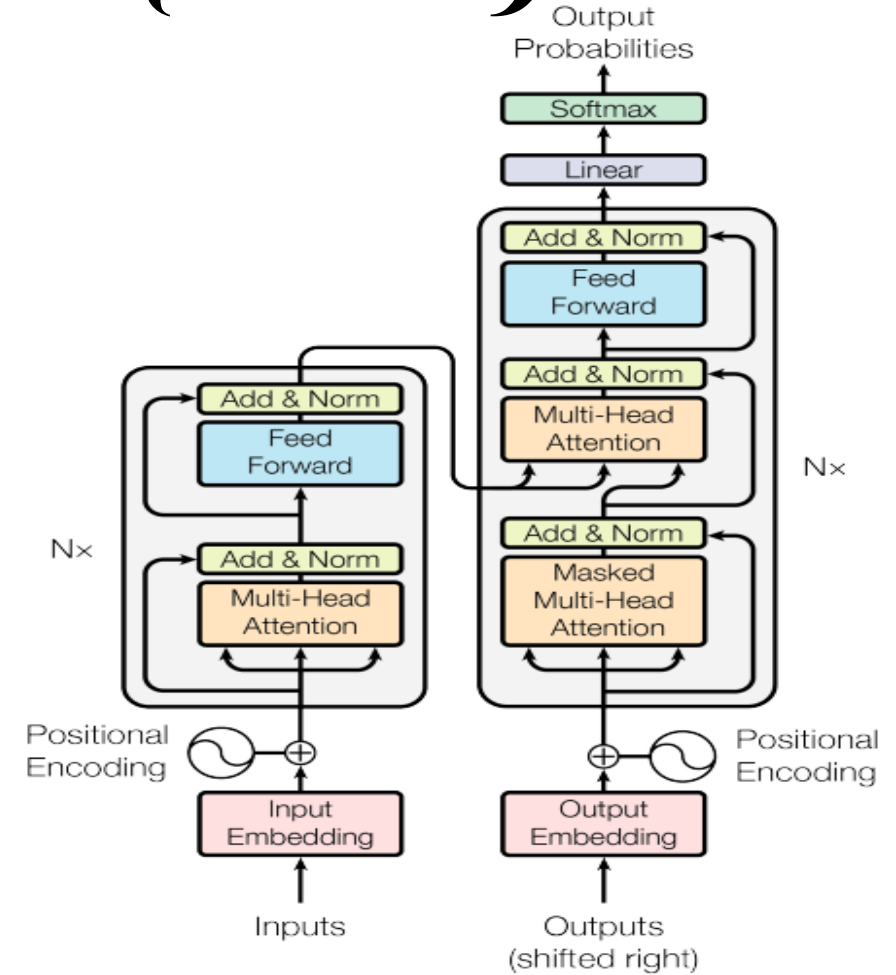


Figure 1. Attention Architecture<sup>1</sup>

# ***Result***



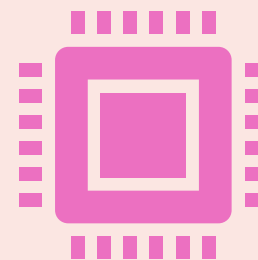
## **Experimental Setup:**

Cross Entropy Loss

Adam optimizer

AutoTokenizer from the  
pretrained "Helsinki-NLP/opus-mt-en-fi"

P100 GPU



## **Hyperparameters:**

Batch Size: 16

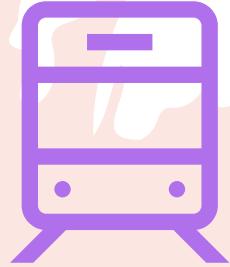
Learning Rate: 0.01

Max Input Length: 512

Max Output Length: 512

Training Epochs: 25

# ***Result***



**Train Loss: 1.8217**



**Test Loss: 1.6819**

The model demonstrates a decreasing trend in both train and test loss.

```
[83]: translate("How are you")  
      तुम कैसे हैं</s>
```

Figure 2. Final Output

# ***Conclusion***



Comprehensive  
Understanding



Effective Utilization



Optimization and  
Fine-Tuning



Practical Application

# ***References***

1. Vaswani, Ashish, et al. "Attention is all you need." *Advances in neural information processing systems* 30 (2017).
2. <https://www.kaggle.com/datasets/preetviradiya/english-hindi-dataset>
3. Watve, Abhinav Y., and Madhuri A. Bhalekar. "English to hindi translation using transformer." 2023 International Conference on Innovative Data Communication Technologies and Application (ICIDCA). IEEE, 2023.

- Thank You

