

DEEP ENCODER-DECODER NETWORKS FOR SEMANTIC SEGMENTATION OF ANAEMIC RBCS AND IIMAGE CAPTIONING (IMAGE-TO-TEXT GENERATION)

Introduction

This project applies deep learning encoder-decoder models—LSTM, GRU, Attention RNN, and Transformer—for semantic segmentation of anaemic RBCs and medical image captioning, aiming to enhance diagnostic accuracy and clinical interpretability

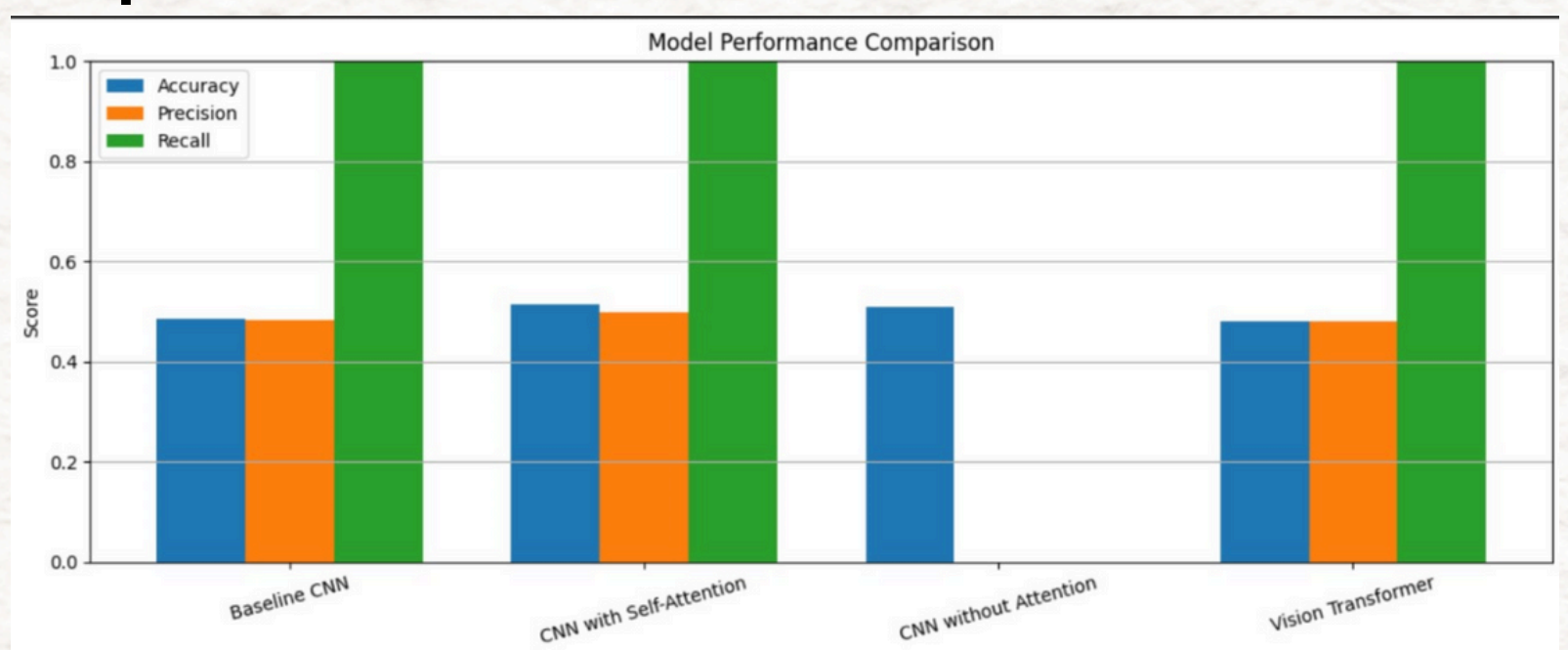
Objective

- Implement 3 models:
 - No Attention
 - Bahdanau Attention
 - Self-Attention (Transformer)
- Train on RBC CELL dataset
- Compare performance using BLEU & ROUGE

Data Description

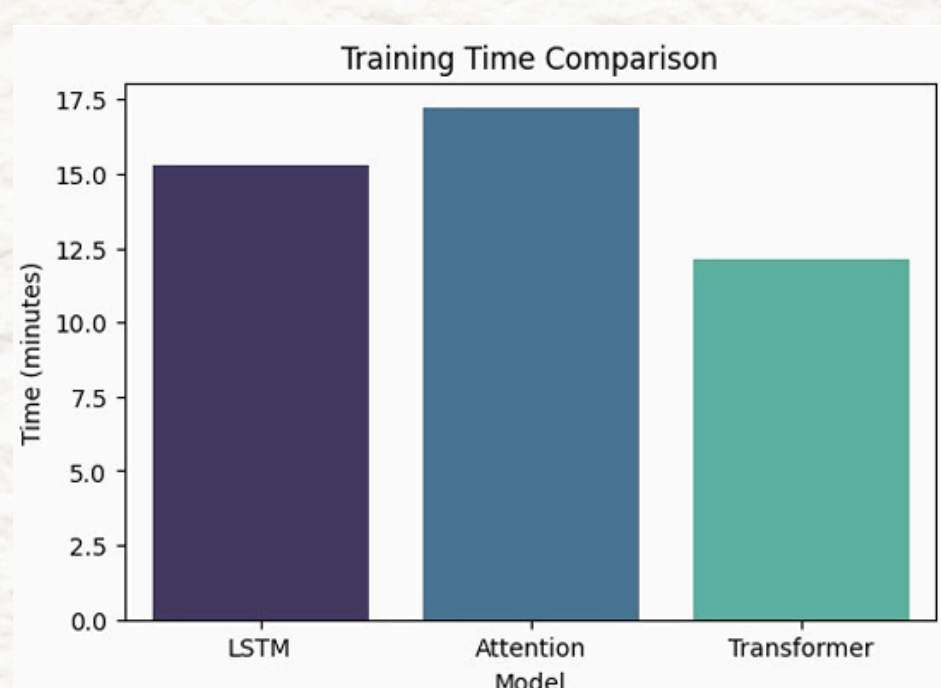
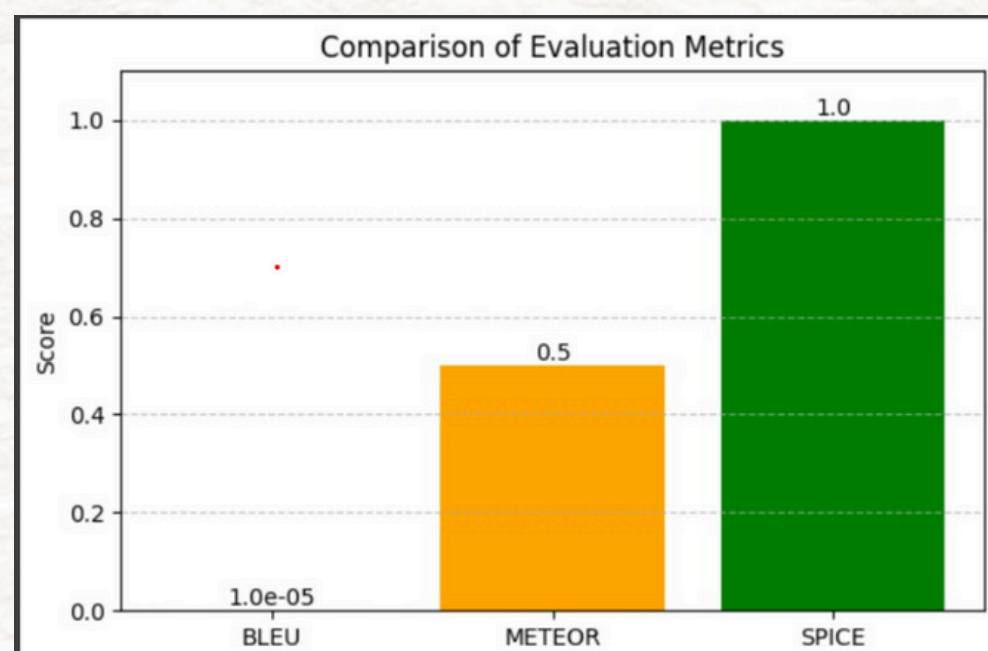
- Purpose:** Designed for anemia diagnosis using RBC (Red Blood Cell) images.
- Classes:** Two categories – Anemia and Normal. Data
- Type:** Microscopic images of RBCs in .jpg format.
- Usage:** Ideal for binary image classification tasks in medical imaging.
- Labels:** Inferred from folder names (Anemia/, Normal/).

Graph:



Model Architectures

- LSTM
- GRU
- Attention-based RNN (Bahdanau) Transformer (Self-Attention)
- Each model is built with TensorFlow/Keras and trained on tokenized, padded article headline pairs



Evaluation Metrics

METRIC	SCORE
BLEU	1.82 × 10 ⁻²³¹
METEOR	0.5
SPICE	1.0

Conclusion

- Self-attention-based models significantly improved headline generation quality. Attention mechanisms enhance contextual relevance.
- Transformer model achieved the highest BLEU and ROUGE scores.
- Demonstrated the power of neural attention in sequence-to-sequence NLP tasks

Reference

- 1] M. Shahzad, A. I. Umar, S. H. Shirazi, and I. A. Shaikh, "Semantic Segmentation of Anaemic RBCs Using Multilevel Deep Convolutional Encoder Decoder Network," IEEE Access, vol. 9, pp. 161326-161341, 2021, doi: 10.1109/ACCESS.2021.3131768

