# ScoreIt-RoBERTa: Scoring clinical patient notes

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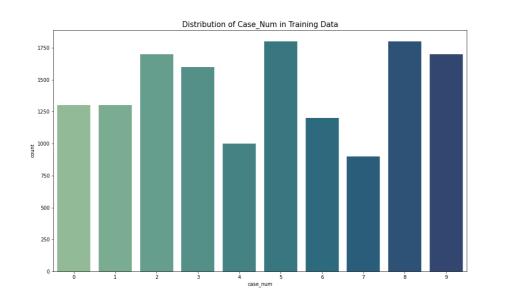
#### Motivation

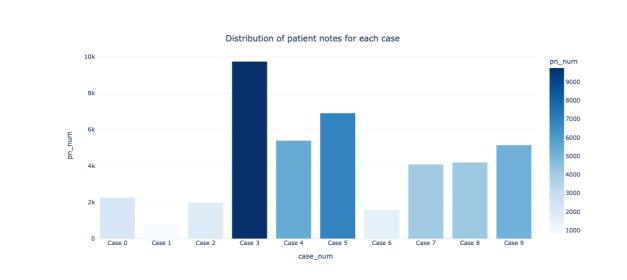
Physicians have a long history of writing patient notes that document history, physical exam findings, possible diagnoses, and follow-up care. Extracting meaningful information from these clinical notes can aid in the accurate diagnosis of diseases.

Named Entity Recognition (NER) is a part of information retrieval and natural language processing that tries to recognize and categorize named entities referenced in unstructured text. The objective of our project is to train a NER model capable of identifying and annotating specific clinical concepts in English patient notes. The developed model can also benefit exams like the United States Medical Licensing Examination (USMLE), where extracted features from the clinical notes can be evaluated against the rubric.

In Medical NER, recognition models based on English clinical patient notes are lacking. We propose designing an English NER model that identifies and annotates the text to find specific clinical concepts based on clinical patient notes.

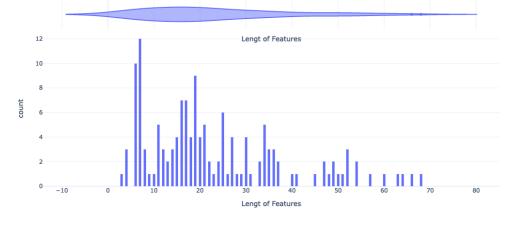
## **Exploratory Data Analysis**

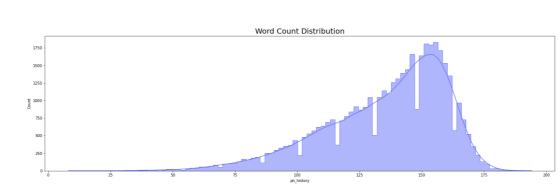




Distribution of Cases in Train Dataset

Distribution of Patient Notes in each case

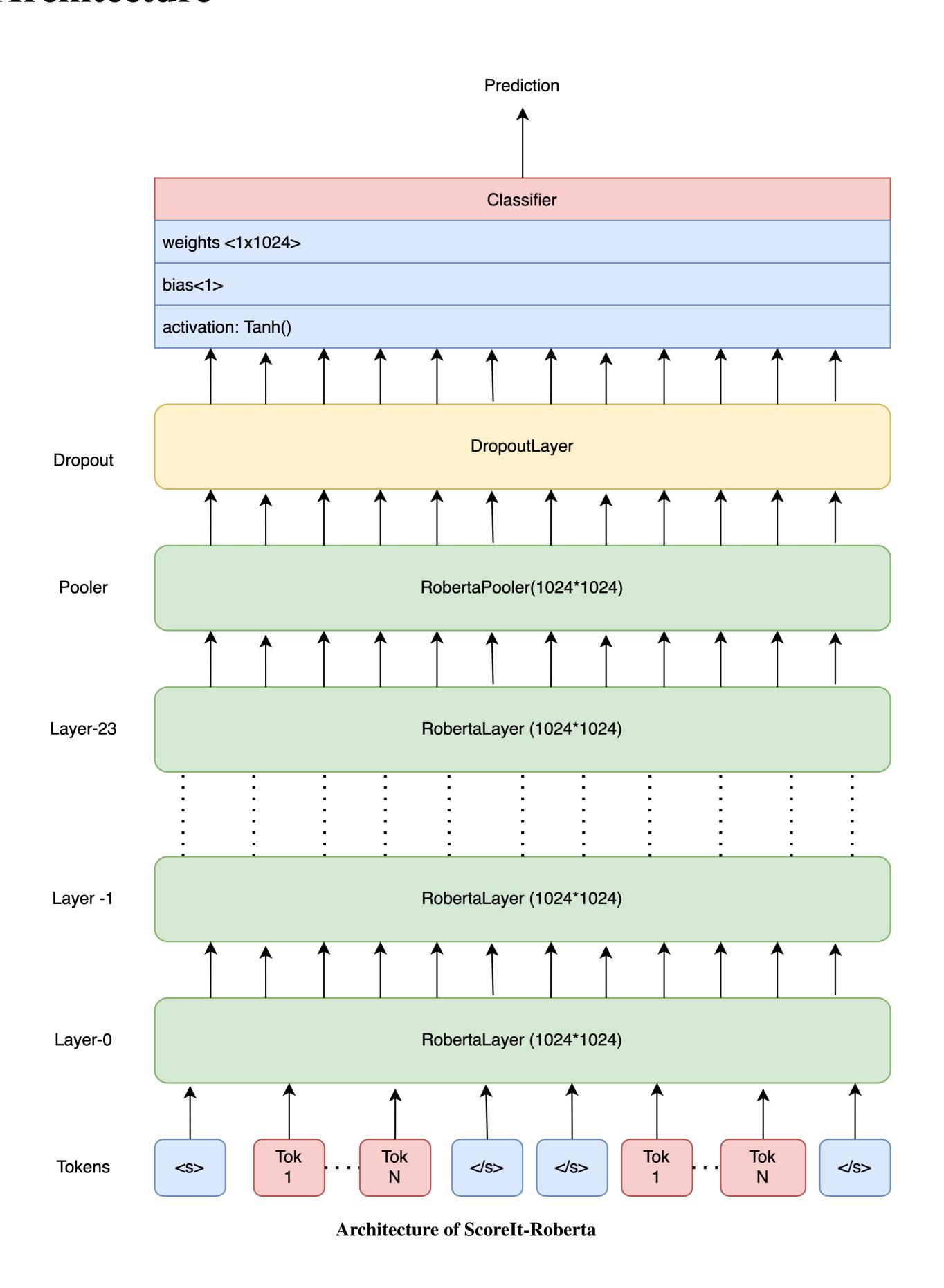




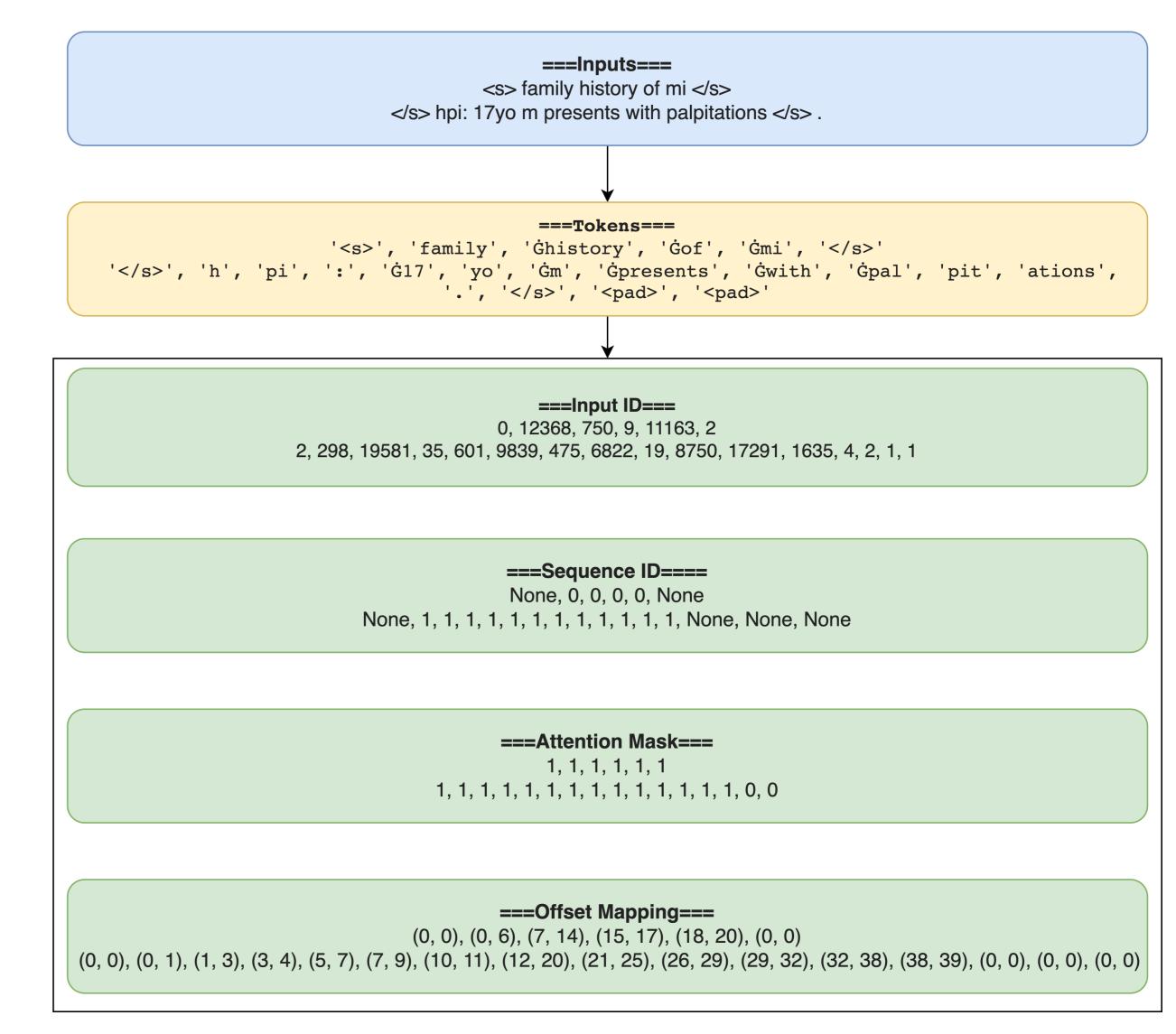
Distribution of Length of Features

Distribution of Word Count

## Architecture



- "RoBERTa-large" model was used for extracting the patient's feature information from the patient history.
- The feature text and the patient history were passed to the tokenizer to get the sequence\_ids, input\_ids and attention mask. These fields are used as an input to our model.
- The RoBERTa Model consists of 23 layers of RoBERTa, followed by a pooled layer, which has the activation function as Tanh.
- A single layer of RoBERTa is divided into sub-layers, which are the attention layer, output layer, intermediate layer and the output layer.
- At the end of the roberta-model, a dropout layer was added. Finally, a linear classifier was used for the prediction.
- The model was trained on over 355 million parameters.



**Tokenization of Sentences** 

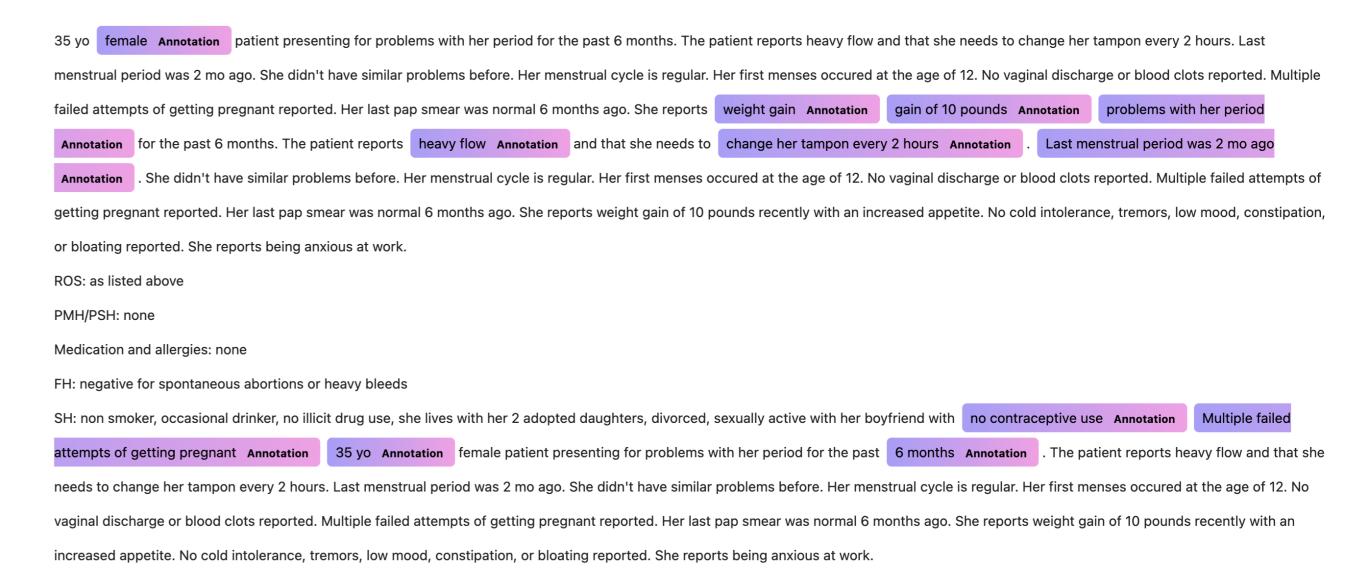
#### **Results and Insights**

We selected three important indicators: precision, recall and F1 score, to evaluate the performance of the named entity recognition task on the dataset presented from the USMLE Step 2 Clinical Skills examination.

RoBERTa was trained for 5 epochs with a stratified k fold split (k=5), yielding the following results.

Metric	Baseline	BERT	RoBERTa
Precision	96.64	75.27	79.89
Recall	12.99	72.34	84.98
F1-Score	22.91	71.27	82.35

Metrics for Baseline, BERT, RoBERTa



Example of Annotation in Patient Notes: Patient Notes highlighted with Annotations

### Acknowledgement

The text data is provided by the National Board of Medical Examiners (NBME) from the USMLE Step 2 Clinical Skills examination, a medical licensure exam.