

NLP Project Group - 20

Topic:- Effective Healthcare answer summarization

A small description of our NLP project, highlighting the key objectives and methodologies.

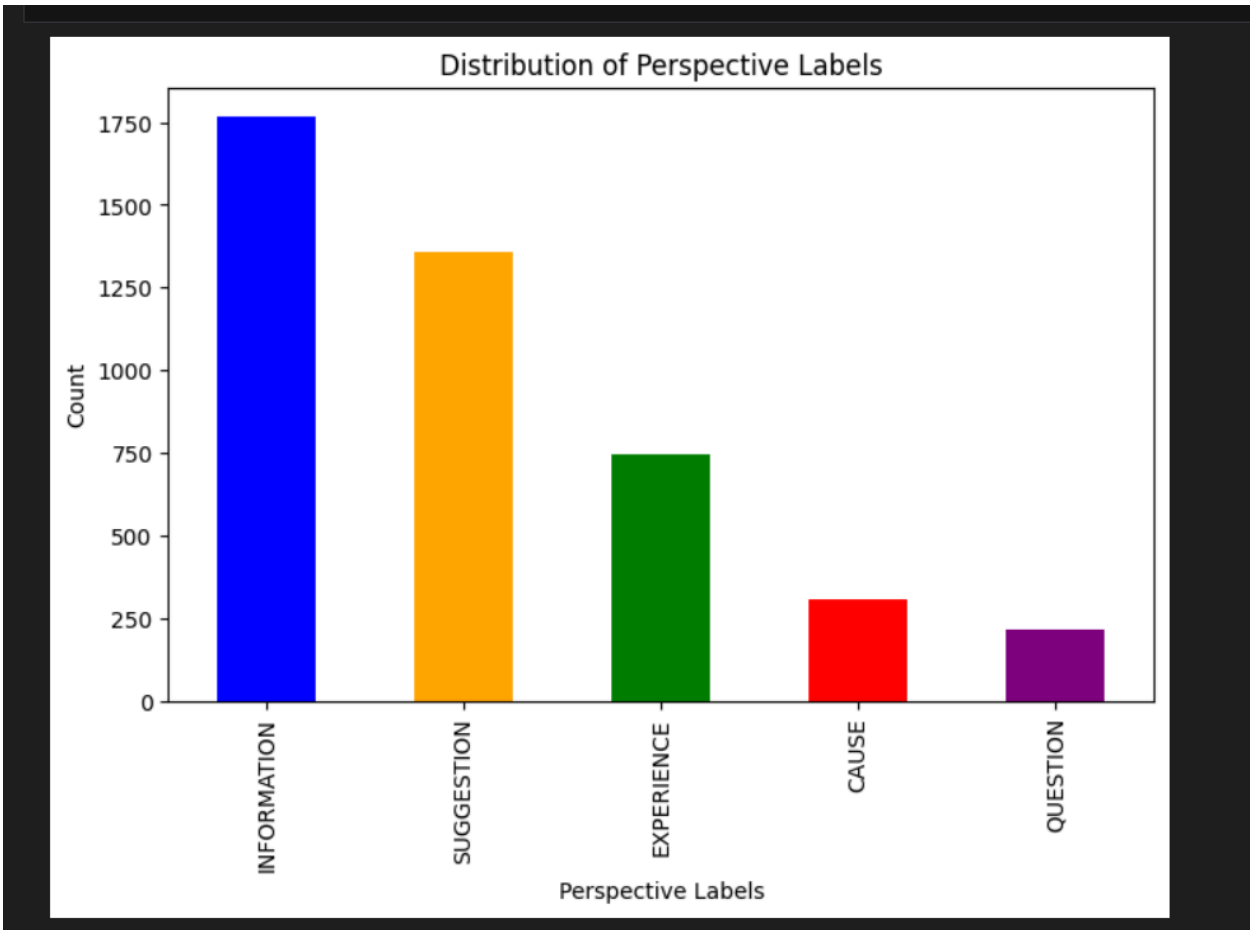
Our project focuses on generating perspective-based summaries of answers by extracting diverse perspectives such as **causes, suggestions, and personal experiences** from multiple user responses to generate a single best summary. For this, we have trained two models:

1. **Flan-T5 Model:** We used the pretrained **Flan-T5** model and further fine-tuned it. Summaries generated by both the pretrained and fine-tuned versions are displayed in the submitted notebook.
2. **Plasma Model:** This model is based on the approach used in a research paper. This model is used for perspective based learning. We have attached the results obtained from this model as well.

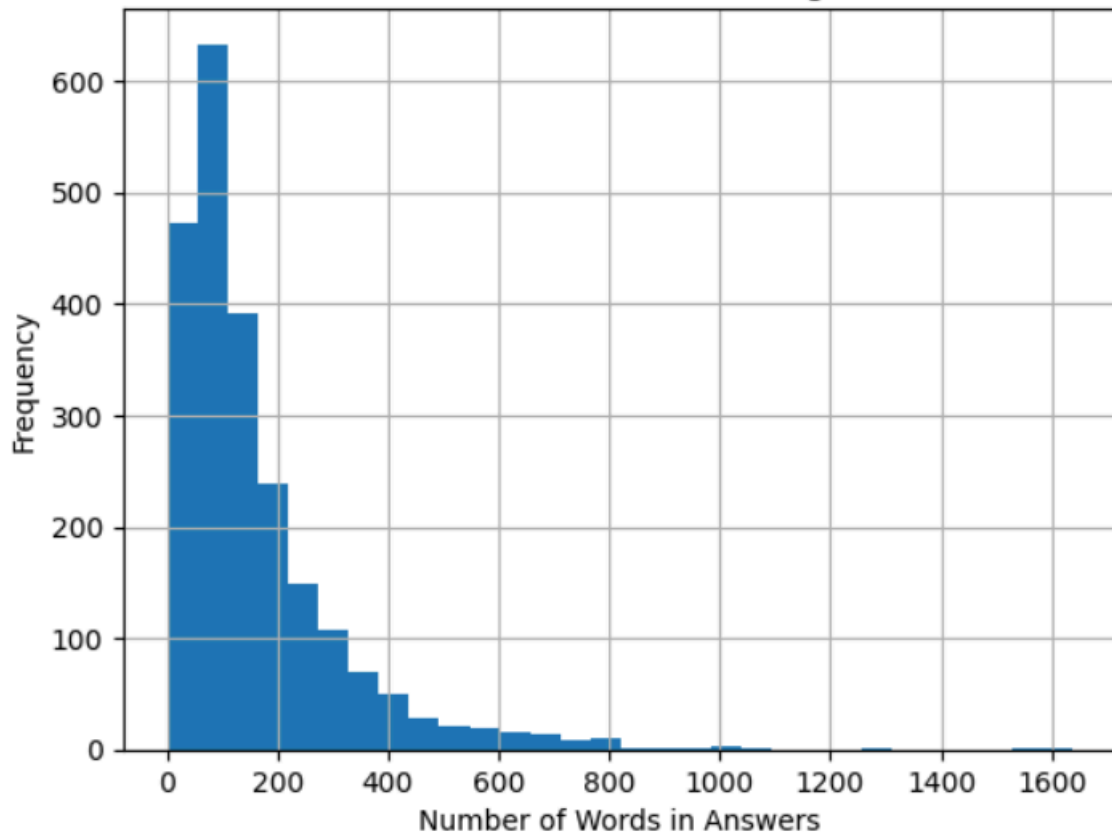
Workflow

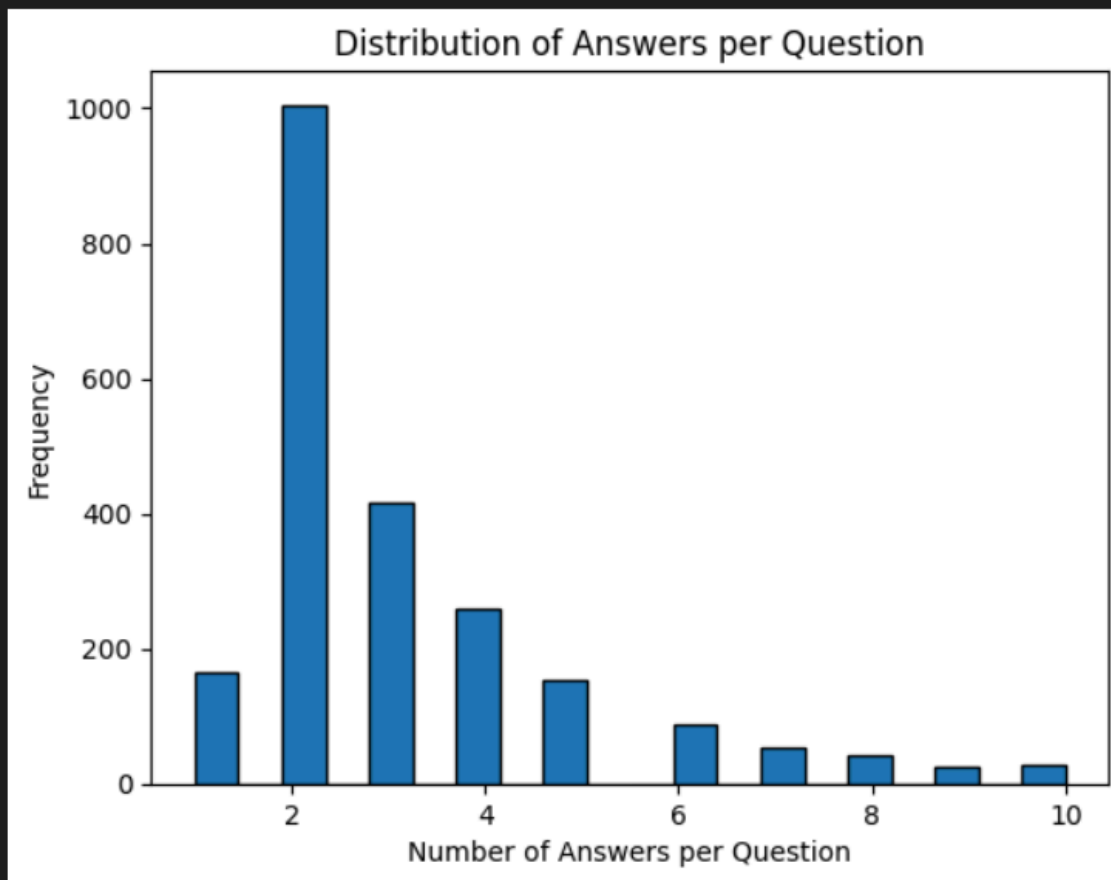
1. Exploratory Data Analysis (EDA):

Performed initial exploration of the dataset (e.g., checking data distribution, missing values, text length).



Distribution of Answer Lengths





2. Data Preprocessing:

- Handled missing values in the dataset.
- Created input-output pairs for training into the required format(question,context,answers,etc)
- Initialized and applied a tokenizer.
- Converted tokenized inputs and outputs into tensors.

3. Model Selection and Fine-Tuning:

- Used the **Flan-T5** model.
- Implemented zero-shot baseline to see out-of-the-box performance
- Fine-tuned the model on the prepared dataset to improve performance

4. Evaluation and Comparison:

- Computed BLEU score for both baseline and fine-tuned models.
- Compared zero-shot vs. fine-tuned results side by side.
- Displayed sample predictions for better qualitative assessment.

For information purposes, an orgasm, also known as a sexual climax, is a pleasurable physical, psychological, or emotional

```
Validation completed. Results saved to './generated/generated_result.csv'
Train Loss: 0.8129 | Validation Loss: 0.8140
Best model saved at ./saved_model
Epoch 5/5
```

Epoch 5

```
Validating: 99% | 238/240 [10:22<00:05, 2.59s/it]Batch 239/240 || Validation Loss: 0.7933
(env) arunb@iitd:~/harshvardhini/project/PUMA-PLASMA-ACL$ ^C
```

Evaluation metrics/ Results

```
(env) arunb@iitd:~/harshvardhini/project/PUMA-PLASMA-ACL$ python3 src/eval.py
ROUGE scores: {'rouge-1': np.float64(0.6907523487404268), 'rouge-2': np.float64(0.08831026839627029), 'rouge-l': np.float64(0.6443098224152211)}
[nltk_data] Downloading package wordnet to /home/arunb/nltk_data...
[nltk_data] Package wordnet is already up-to-date!
[nltk_data] Downloading package punkt_tab to /home/arunb/nltk_data...
[nltk_data] Package punkt_tab is already up-to-date!
[nltk_data] Downloading package omw-1.4 to /home/arunb/nltk_data...
[nltk_data] Package omw-1.4 is already up-to-date!
METEOR score: {'meteor': np.float64(5.861975132291988)}
BLEU scores: {'bleu-1': np.float64(3.3225238069990426), 'bleu-2': np.float64(1.26544745564501), 'bleu-3': np.float64(0.74477327350868), 'bleu-4': np.float64(0.4649751114406014)}
Some weights of RobertaModel were not initialized from the model checkpoint at roberta-large and are newly initialized: ['pooler.dense.bias', 'pooler.dense.weight']
You should probably TRAIN this model on a down-stream task to be able to use it for predictions and inference.
Warning: Empty candidate sentence detected; setting raw BERTscores to 0.
Warning: Empty candidate sentence detected; setting raw BERTscores to 0.
BERTScore: {'bertscore': 75.97826719284058}
```

Next Steps and Improvements

So far, we have implemented a basic approach by referencing the research paper, training two models, and evaluating their performance using simple preprocessing techniques. Moving forward, our goal is to enhance the performance of the model by:

1. **Researching Advanced Models:** Exploring state-of-the-art models specifically designed for medical question answering to improve summarization quality.
2. **Stacking Models:** Implementing a combination of models to leverage different strengths and generate better results.
3. **Fine-tuning with Diverse Pretrained Models:** Experimenting with multiple pretrained models to compare their effectiveness in our task.
4. **Optimizing Preprocessing:** Improving data preprocessing techniques to better prepare the input for model training.
5. **Hyperparameter Tuning:** Experimenting with different hyperparameter settings to optimize model performance.
6. **Incorporating External Datasets:** If applicable, expanding our dataset to improve generalization and robustness.

Execution Instructions

Running the Project

This project was executed using **Google Colab** for training the **Flan-T5** model and a **GPU server from our college** for training the **Plasma Model**. To run the project:

1. **Open the Jupyter Notebook ([NLP_Project_20.ipynb](#))** in Google Colab.

2. **Install required libraries** by running the necessary setup cells at the beginning of the notebook.
3. **Follow the execution order** and run each cell sequentially to replicate our results.
4. For the **Plasma Model**, refer to the precomputed results attached in the notebook, as it was trained on the college GPU server.