

Infosys Internship 4.0 Project Presentation

Title:TEXT SUMMARIZATION

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CSE(Internet Of Things With Cyber Security And Blockchain
Technology)

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❖ **Acknowledgment**

- I would like to thank Infosys for giving opportunity to work as an AI/ML Intern.
- Special thanks to my mentor, Mr. Narendra Kumar, for his invaluable guidance and support.
- I am also grateful to my team and colleagues for their continuous support and collaboration throughout this project.
- Additionally, I am thankful to my family for their encouragement and assistance throughout this project.

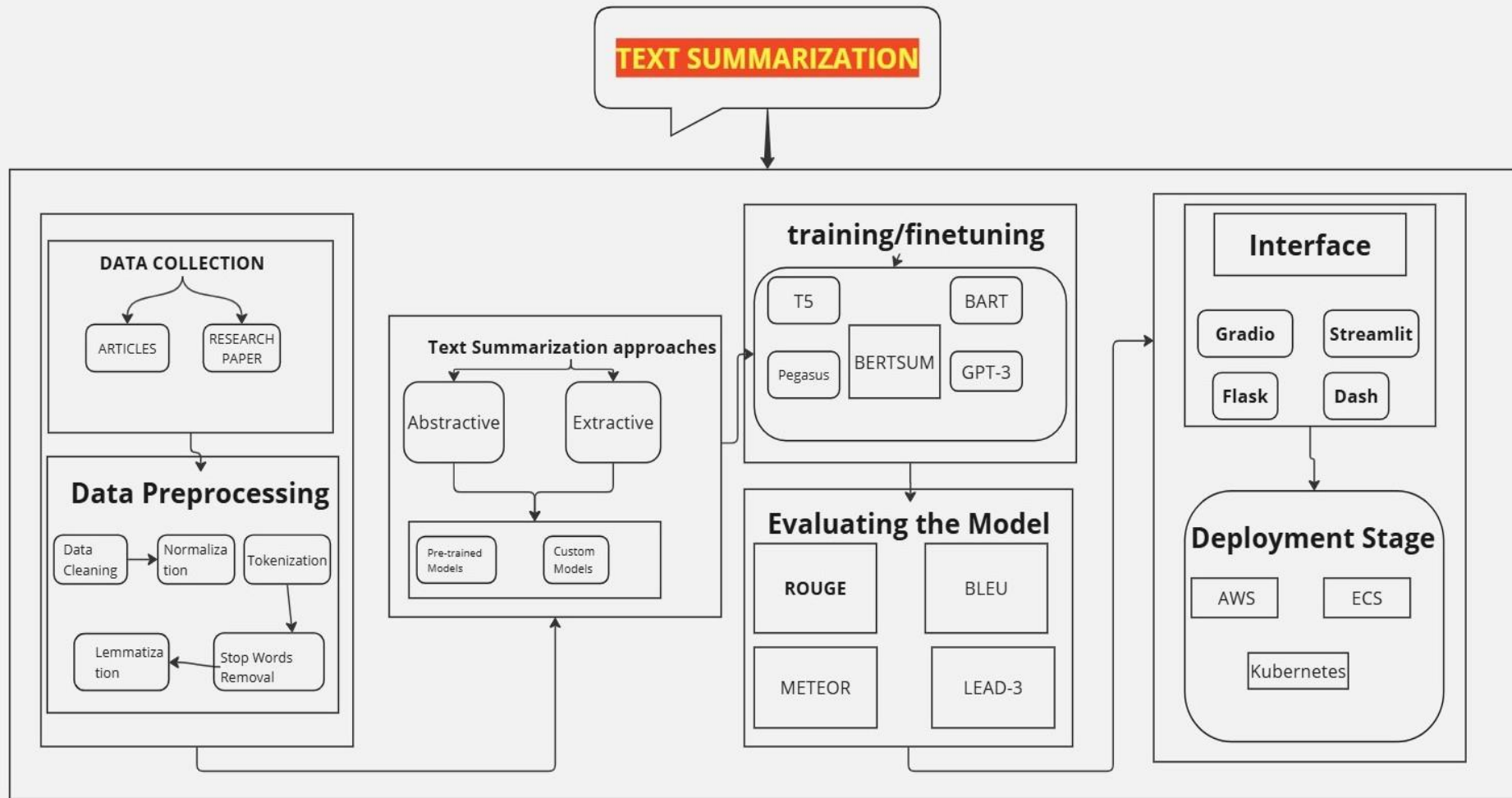
❖ **Problem Statement**

The project aims to develop an effective text summarization model that condenses extensive texts into concise summaries with high accuracy and relevance.

❖ **Introduction**

This report outlines the problem statement, workflow, data collection, preprocessing steps, summarization methodologies, user interface design, and results. It also discusses observations, conclusions, and potential future enhancements, showcasing the system's ability to improve information retrieval and readability across various domains.

❖ workflow



❖ Data collection

Source: I collected a dataset from the CNN/DailyMail website for this project.

Relevance: This dataset is ideal for training summarization models because of its rich and diverse content.

Dataset Description

- **Initial Dataset:** The main dataset was initially saved as dataset.csv, containing 70,000 records.

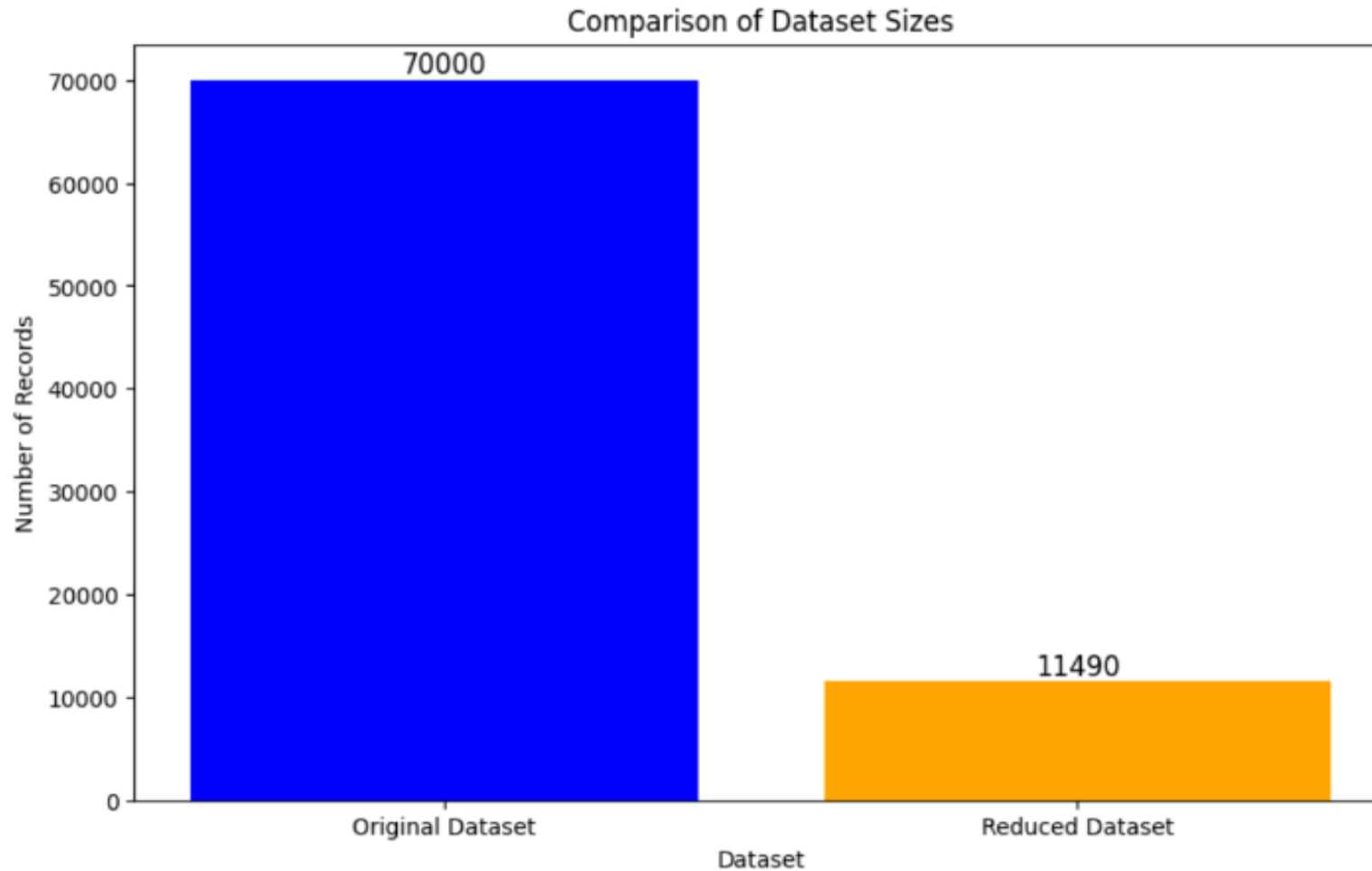
Structure:

- **id:** Unique identifier for each record.
- **article:** The body of the news article.
- **highlights:** The summary or highlights of the article.

Dataset Links

- **Master Dataset:** You can find the link to the master dataset [here](#).

Data Reduction



- Reduced Dataset: You can find the link to the reduced dataset [reduced data](#).

❖ Data Preprocessing

Objective:

- The primary goal of data preprocessing was to clean and prepare the dataset for model training by removing noise, tokenizing the text, and eliminating stopwords.

Process:

- The dataset before preprocessing is shown in the below picture

Out[2]:

	id	article	highlights
0	8aa8d3d042356a88d25ee6fb13347184858fe770	(RollingStone.com) -- Britney Spears announce...	Britney Spears and producers still choosing so...
1	b3a6c45ccbcc6140a9fe042a385440e3a80535dc	By . Sam Adams . PUBLISHED: . 04:02 EST, 18 Ju...	Car owners would be liable even if they don't ...
2	f90015991bcec3013e502044699046581088f1a5	It is a single moment of horrifying barbarism ...	The picture was posted on a pro-government web...
3	0e029a3f67dc8df34eefc185ec5343cec72fb29d	An elderly Minnesota couple were killed after ...	Carlton and Hazel Roed of Mentor, Minnesota, w...
4	b244323ba60a10baf71a72a30ffed5162f3b2050	(CNN) -- Columbus Day often brings to mind the...	Seattle and Minneapolis will celebrate Indigen...

Outcomes:

- Below picture shows cleaned dataset after preprocessing

Steps Undertaken:

- Cleaning Text Data:**
- Tokenization:**
- Stopword Removal:**
- Preprocessing Articles and Highlights:**
- Dataset Splitting:**

[preprocessing](#)

Out[8]:

	id	article	highlight
0	8aa8d3d042356a88d25ee6fb13347184858fe770	rollingstonecom britney spears announced today...	britney spears producers still choosing songs ...
1	b3a6c45ccbcc6140a9fe042a385440e3a80535dc	sam adams published 0402 est 18 july 2012 upda...	car owners would liable even dont know dropped...
2	f90015991bcec3013e502044699046581088f1a5	single moment horrifying barbarism provides fl...	picture posted progovernment website lebanon b...
3	0e029a3f67dc8df34eefc185ec5343cec72fb29d	elderly minnesota couple killed car collided h...	carlton hazel roed mentor minnesota 2009 chevy...
4	b244323ba60a10baf71a72a30ffed5162f3b2050	cnn columbus day often brings mind nina pinta ...	seattle minneapolis celebrate indigenous peopl...

Dataset Splitting

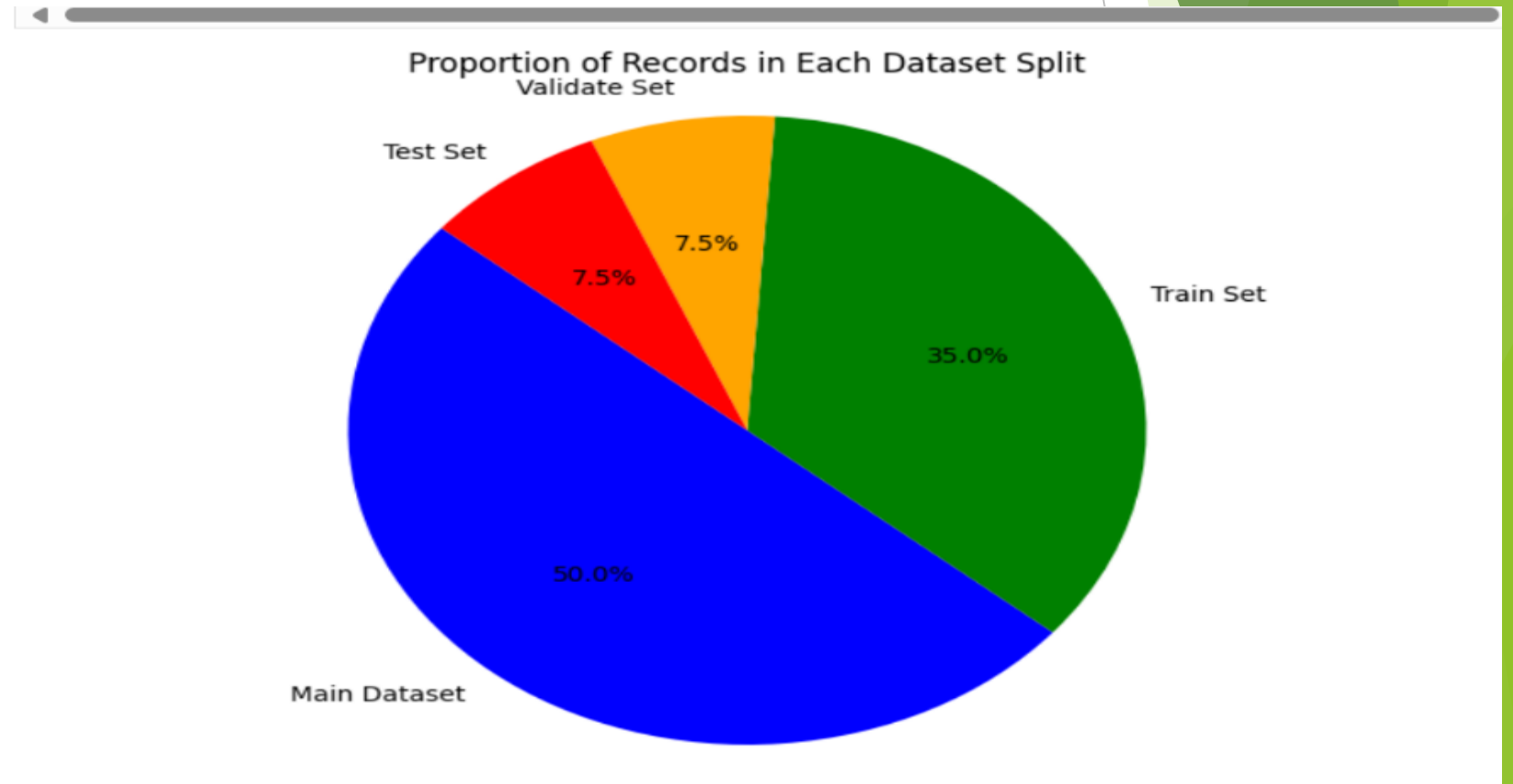
- Split the preprocessed dataset into three distinct sets: training (70%), validation (15%), and test (15%) sets to facilitate model training and evaluation.
- A pie chart was generated to visually represent the proportion of records in each dataset split

- Splitted Dataset links

[train_data](#)

[validate_data](#)

[test_data](#)



❖ Abstractive Summarization

- **Definition:** Abstractive summarization creates new sentences to capture the essence of the original text.
- **Examples:** T5, BART, and GPT.
- **Selected model:** T5-small Transformer model from Hugging Face
- **Training process:**

Initial Losses: Training loss started at 1.05 and validation loss at 0.96

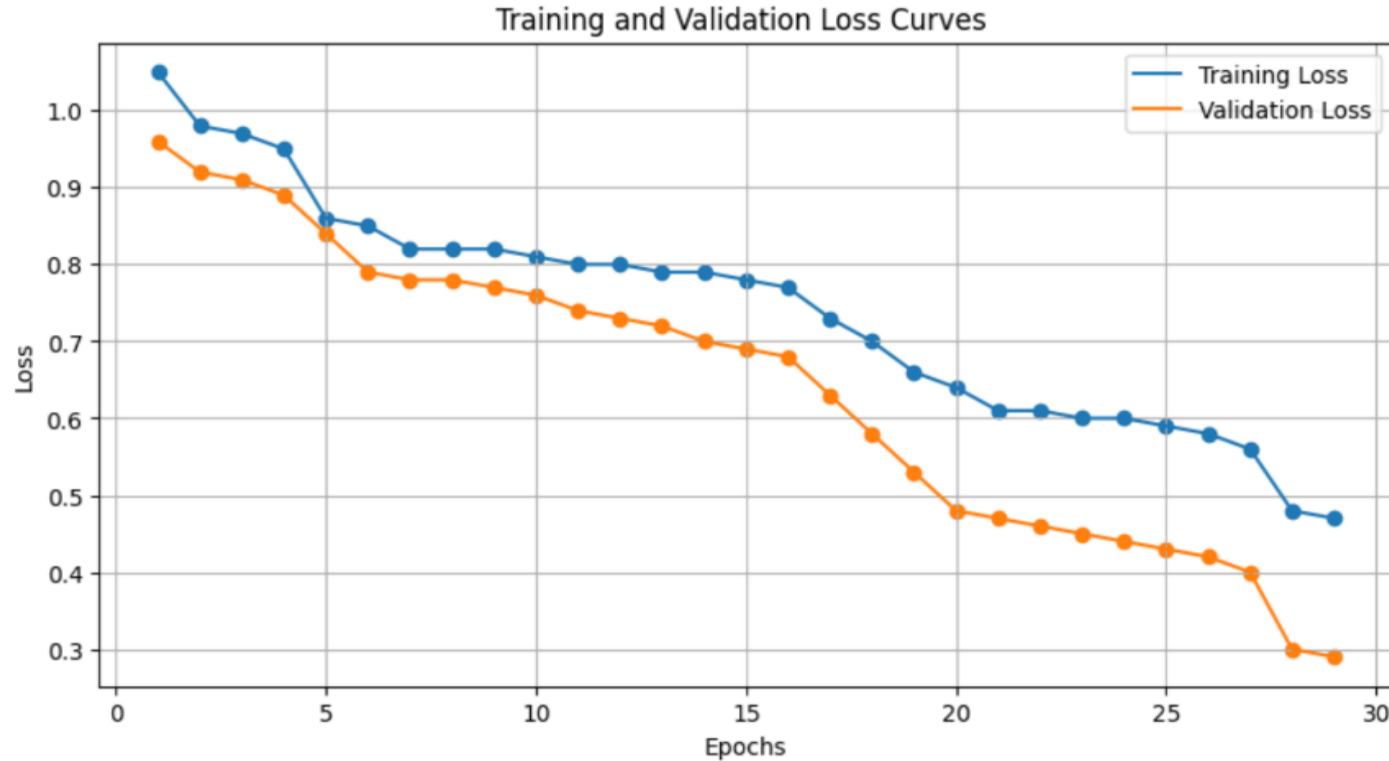
```
warnings.warn(
Epoch 1: 100%|███████████| 403/403 [10:07:02<00:00, 90.38s/it, train_loss=1.25]
Average training loss: 1.0562968504044317
Validation loss: 0.9614474930982481
Further training completed and model saved to fine_tuning.
```

Final Losses: Reduced to 0.47 for training and 0.29 for validation.

```
Average training loss: 0.47855778578759567
Validation loss: 0.2974807008135098
Model improved. Saving the model.
Training completed.
```

● Observations:

Loss Reduction Post Hyperparameter Tuning



- The validation loss saw a significant decrease from an initial 0.96 to a final 0.29 after hyperparameter tuning
- Simultaneously, the training loss also reduced notably from 1.05 to 0.40, indicating improved model convergence and effectiveness in generating accurate summaries

❖ Performance Matrix

Objective:

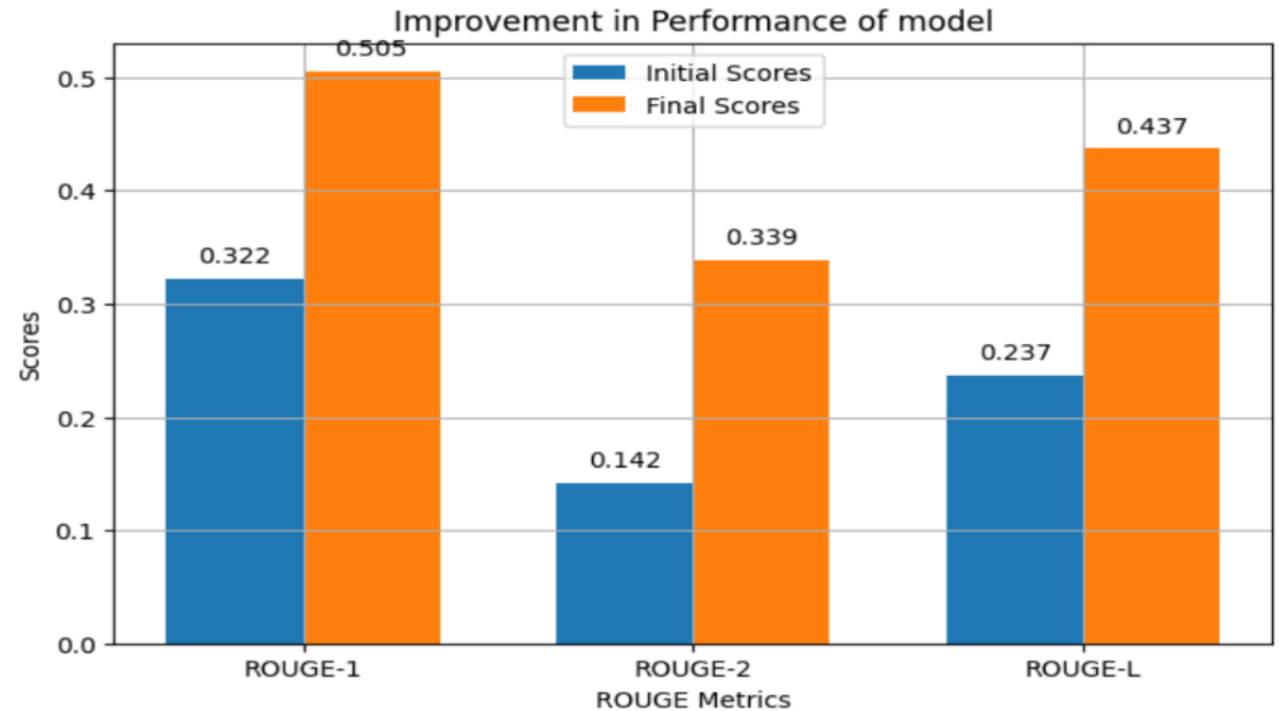
- **Evaluate Model Performance:** Quantitatively assess how well the generated summaries match the reference summaries.
- **Available Evaluation Metrics:** BLEU, ROUGE,
- **Used Performance Matrix (ROUGE):** Recall-Oriented Understudy for Gisting Evaluation
- **ROUGE Score Analysis:**
 - ROUGE-1: ▪ ROUGE-2: ▪ ROUGE-L:

```
Successfully installed rouge score 0.1.2
Special tokens have been added in the vocabulary, making
100%|██████████| 216/216 [06:21<00:00, 1.76s/it]
ROUGE-1: 0.5051 ROUGE-2: 0.3398 ROUGE-L: 0.4377
```

❖ Observations

Performance Improvement

- Post tuning, these metrics improved to 0.505, 0.339, and 0.437, demonstrating substantial progress in summarization accuracy
- Initial ROUGE-1, ROUGE-2, and ROUGE-L scores stood at 0.322, 0.142, and 0.237 respectively



- <https://github.com/Nishashetty664/text-summarization-project/blob/16a2fd1e635a96fae941fc8d7643d2d3e834bfe7/milestone1/final%20training%20code.ipynb>
- https://github.com/Nishashetty664/text-summarization-project/blob/16a2fd1e635a96fae941fc8d7643d2d3e834bfe7/milestone1/evaluation_abstractive.ipynb

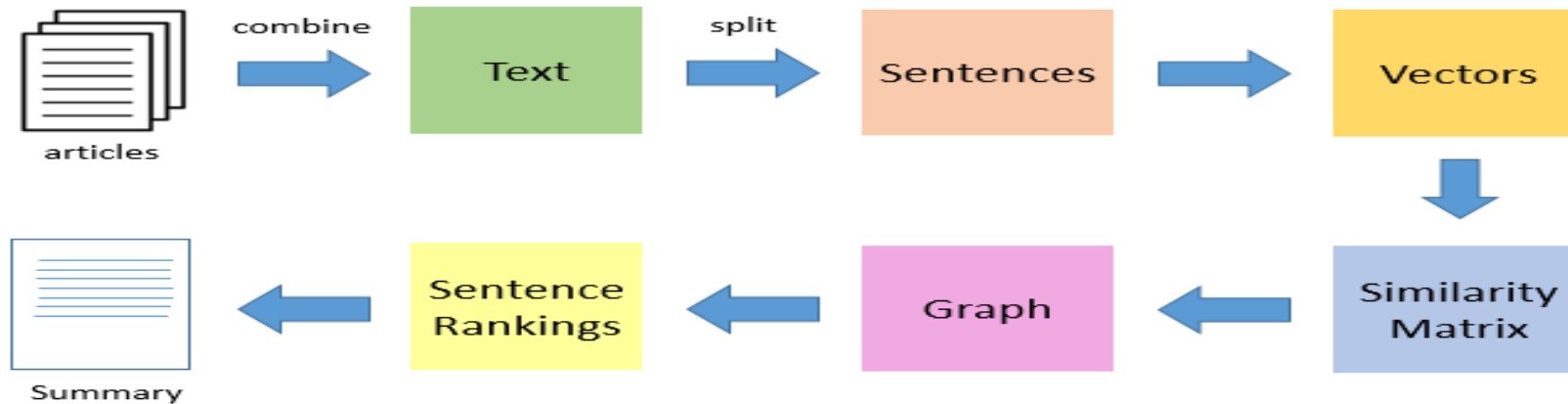
❖ Extractive text summarization

- **Objective:** The objective is to develop a system that generates concise summaries by selecting key sentences from the source text, and preserving original wording and context.
- **Examples:** Techniques like TextRank, TF-IDF, and pre-trained models such as BERT and GPT.

Model training methodology

- Used textrank algorithm

Methodology



• Performance Metrics and Scores:

- Used Performance Matrix (ROUGE):
 - **ROUGE Score** (Recall-Oriented Understudy for Gisting Evaluation).
- **RESULT:**
 - **Precision:** The average precision score was approximately 0.119.
 - **Recall:** The average recall score was approximately 0.813.
 - **F1 Score:** The average F1 score was approximately 0.203

```
Summarization complete. Summarized data saved to 'summarized_train_new1.csv'.  
Average Precision: 0.11943901651820747  
Average Recall: 0.8127170619817642  
Average F1 Score: 0.20283625093778002
```


- **Observations**

- **High Recall:** High recall indicates that the model captures most of the relevant information from the original text.
- **Moderate Precision and F1 Scores:** The model shows moderate precision (0.119) and a balanced F1 score (0.203), indicating the potential for improving sentence relevance selection in summaries.
- **Resource Efficiency:** The model is resource-efficient and suitable for environments with limited computational power.
- training and evaluation

❖ User Interface

- **Overview**

- The interface is developed using the Gradio library, providing an interactive platform for text summarization.
- Users can input text or upload PDF files and choose between abstractive and extractive summarization methods.

- **Implementation Overview**

Uses the T5 model to generate abstractive summarization

Implements the TextRank algorithm using CountVectorizer, cosine similarity, and PageRank to rank and select sentences for the extractive summarization

PDF Text Extraction: Uses [PyMuPDF \(fitz\)](#) to extract text from PDF files.

[interface_link](#)

❖ Results

The screenshot displays the 'TEXT SUMMIFY' web application interface. At the top, there are tabs for 'Summarize Text' and 'About TEXT SUMMIFY'. The main heading 'TEXT SUMMIFY' is prominently displayed. Below this, the 'Input Type' section has two radio buttons: 'Text' (selected) and 'PDF'. The 'Summarization Type' dropdown menu is open, showing 'Abstractive' as the selected option, with 'Extractive' as an alternative. The 'Textbox' contains a long paragraph about a teacher, Madeline Luciano, who was fired for allegedly encouraging bullying. Below the textbox are 'Submit' and 'Clear' buttons. The 'Summary' section at the bottom shows a condensed version of the input text. At the very bottom, there are links to 'Use via API' and 'Built with Gradio'.

Summarize Text About TEXT SUMMIFY

TEXT SUMMIFY

Input Type

☒ Text ☐ PDF

Summarization Type

Abstractive

✓ Abstractive

Extractive

Textbox

Madeline Luciano, a 40-year-old teacher at PS 18 in Manhattan, found herself at the center of a controversy when she was fired for allegedly encouraging her eighth-grade pupils to bully a 13-year-old girl by having them write the girl's worst qualities on the blackboard. The incident, which took place last June, saw the students write hurtful words such as "ugly," "annoying," and "phony" to describe their classmate, who had been a target of bullying multiple times before. The situation escalated when the girl, visibly upset and distressed, began to cry in response to the cruel comments written about her. Ms. Luciano, who had been employed by the New York education department since 2010, claims that her intentions were misunderstood. She insists that her aim was to teach the students about the harmful effects of bullying and never intended for the situation to spiral out of control. According to Ms. Luciano, the exercise was meant to be an educational tool to illustrate the evils of bullying, but it unfortunately took a negative turn. In the aftermath of the incident, Ms. Luciano has launched a court action against the city's education department in an effort to regain her teaching license and return to her profession. She maintains that she has always employed various strategies to address and modify her students' behavior, and the unfortunate outcome of this particular incident does not reflect her overall approach to teaching and discipline. Ms. Luciano's case highlights the complexities and challenges educators face in addressing bullying within the classroom, and her fight to clear her name continues as she seeks justice and the opportunity to resume her teaching career.

Submit

Clear

Summary

madeline Luciano 40 fired accidentally encouraging eighth grade pupils to bully a 13yearold girl whose qualities on the blackboard became visibly upset distressed began to cry in response cruel comments

Use via API · Built with Gradio

- The above figure shows the interface that summarizes long text into a summary using abstractive summarization with the T5-small model

❖ Results

The screenshot displays the TextRANK web application interface. At the top, there are two radio buttons for 'Text' and 'PDF', with 'PDF' being selected. To the right of these is a dropdown menu currently set to 'Extractive'. Below this is a file upload section with a button labeled 'Upload PDF File' and a close icon. A file named 'test.pdf' is shown with a size of '70.2 KB' and a download icon. Below the upload section are two large buttons: 'Submit' and 'Clear'. The bottom section, titled 'Summary', contains the following text:

Article 44 - Uniform civil code for the citizens
Context - The Supreme Court in a case concerning the question of whether succession and inheritance of a Goan domicile is governed by the Portuguese Civil Code, 1867 or the Indian Succession Act of 1925, held that the Constitution in Article 44 requires the State to strive to secure for its citizens a Uniform Civil Code (UCC) throughout India, but till date, no action has been taken in this regard Article 39-A of the Constitution directs the State to ensure that the operation of the legal system promotes justice on a basis of equal opportunity and shall, in particular, provide free legal aid by suitable legislation or schemes or in any other way The court held that making NOTA applicable in Rajya Sabha elections is contrary to Article 80(4) of the constitution and the Supreme Court's judgment in PUCL v Union of India (2013)

The above figure shows the interface that summarizes by extracting the text from the pdf using textrank algorithm.

- **Observations**

- Efficient performance in summarizing both text and PDF inputs.
- It produced accurate summaries using the T5 model for abstractive summarization and TextRank for extractive summarization.
- Summarized outputs were concise and preserved essential information from the original texts or documents.

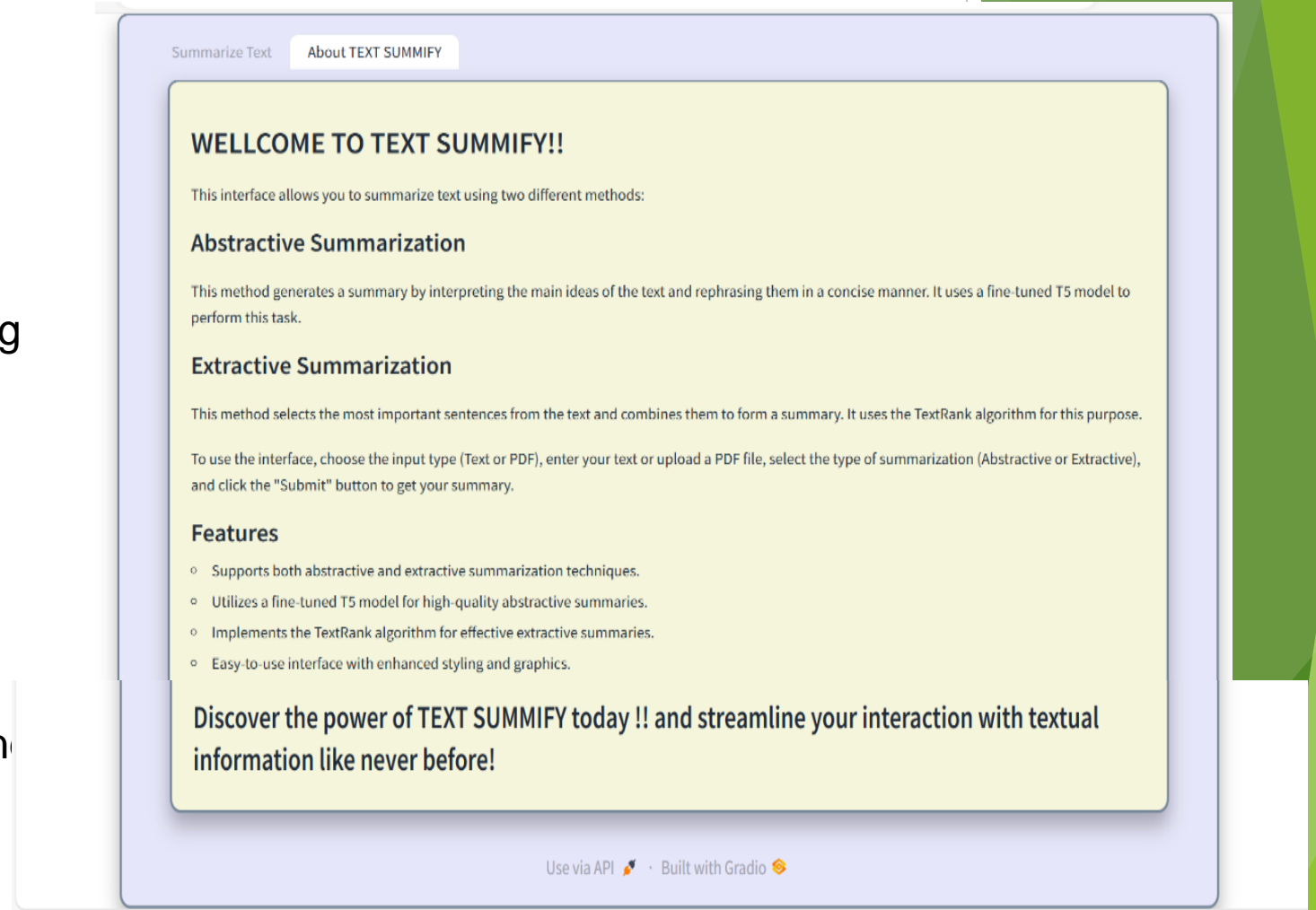


Fig:Description page

❖ Challenges Faced and Solutions

Challenge: Initially faced difficulties in improving model performance due to resource limitations and increased training times on local systems.

Solution: Overcame these limitations by migrating computations to Google Colab, utilizing powerful GPUs to expedite training and inference processes. Additionally, downsized the dataset through strategic sampling and preprocessing techniques to optimize efficiency without compromising quality.

❖ Conclusion

- The development of an automated text summarization system using advanced NLP techniques has proven effective and efficient.
- The Gradio interface provided a user-friendly platform for summarizing long texts and PDF documents.
- This project highlighted the potential of NLP technologies to enhance information retrieval and readability, benefiting various business applications and beyond.

❖ **Future scope**

- **Model Improvement:** Implementing more advanced pre-trained models such as T5-base or T5-large could further improve the accuracy and coherence of the summaries.
- **Multi-language Support:** Expanding the system to support multiple languages to cater to a more diverse user base.
- **Additional Summarization Techniques:** Exploring and incorporating other summarization methods, such as neural network-based extractive summarization and hybrid models, to offer users more choices.
- **Integration with Other Platforms:** Integrating the summarization tool with popular content management systems and collaboration platforms to broaden its applicability.

Thank you 🌟