TEXT SUMMARIZATION PROJECT

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Introduction

Text summarization is the process of condensing a long text into a shorter version while retaining its key information and meaning.

Importance:

- Saves time by extracting relevant information quickly.
- Enhances comprehension by focusing on key points.

Applications:

- News summarization
- Legal document summarization
- Academic paper abstracts
- Customer feedback analysis
- Chatbot and virtual assistant responses



Problem Statement

- Developing an automated text summarization system that can accurately and efficiently condense large bodies of text into concise summaries is essential for enhancing business operations.
- This project aims to deploy NLP techniques to create a robust text summarization tool capable of handling various types of documents across different domains.
- **Objective:** The system should deliver high-quality summaries that retain the core information and contextual meaning of the original text.

Methodology Overview



Abstractive Summarization

Involves generating a summary by paraphrasing the original text.

It uses advanced natural language processing (NLP) techniques to understand the content and create new sentences that convey the core meaning of the source material.

Generates novel sentences that may not appear in the original text.

Model Used: BART



Extractive Summarization

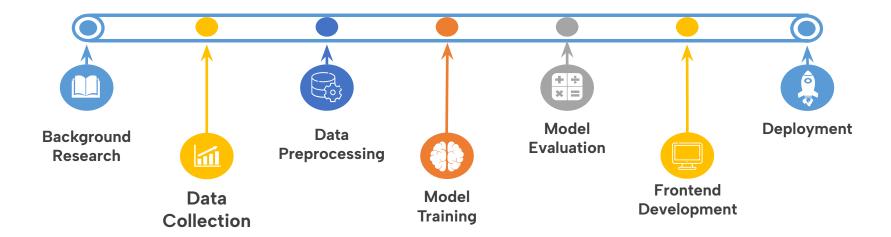
Involves selecting and extracting key sentences or phrases directly from the source text to create a summary.

It does not generate new sentences but instead identifies the most important content and reuses it as-is.

The summarized text retains the original wording and structure of the source.

Model Used: BERT

Workflow Of The Project





Dataset Description

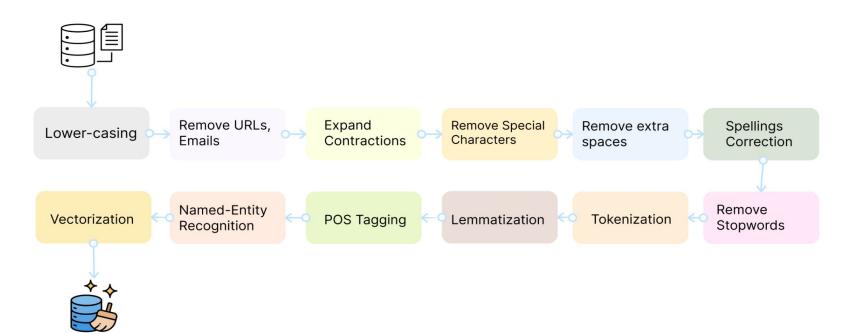
- Dataset: CNN-DailyMail News Text Summarization Dataset
- The CNN DailyMail Dataset is an English-language dataset containing just over 300k unique news articles as written by journalists at CNN and the Daily Mail.



CNN-DailyMail News Text Summarization

News Articles and summary from CNN-DailyMail Dataset

Preprocessing Steps:





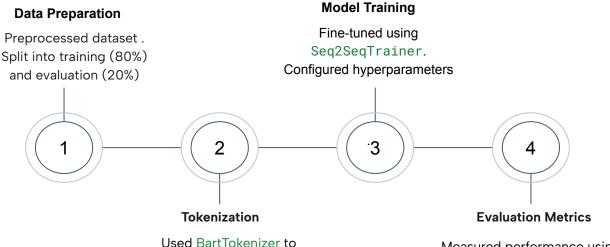
Abstractive Summarization

Fine-tuned BART-large-CNN for abstractive text summarization.

process text.

Applied truncation for

uniform input size.



Results:

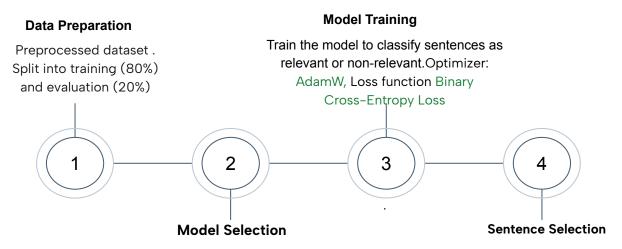
- Training Loss:
 Decreased
 consistently.
- Validation Loss: Early stopping avoided overfitting.
- ROUGE Scores: Improved across epochs.

Measured performance using ROUGE-1, ROUGE-2, and ROUGE-1



Extractive Summarization

Extract key sentences directly from the text to form summaries.



Used BERT Based extractive

summarizer Lightweight transformer optimized for speed and performance. Fine-tuned to score sentence importance.

Sentences are ranked based on importance scores.

Top sentences are selected to form a coherent summary.

Evaluation Metrics

- ROUGE (Recall-Oriented Understudy for Gisting Evaluation) is a set of metrics commonly used to
 evaluate the quality of summaries by comparing the overlap between the generated summary and a
 reference summary.
- Measures the similarity of n-grams (sequence of words) between the generated and reference summaries.

1. ROUGE-1

- Measures overlap of unigrams (single words) between the generated and reference summaries.
- **Significance:** Basic word-level match, indicating content preservation.

2. ROUGE-2

- Measures overlap of bigrams (pairs of consecutive words) between the summaries.
- **Significance:** Reflects fluency and contextual relevance.

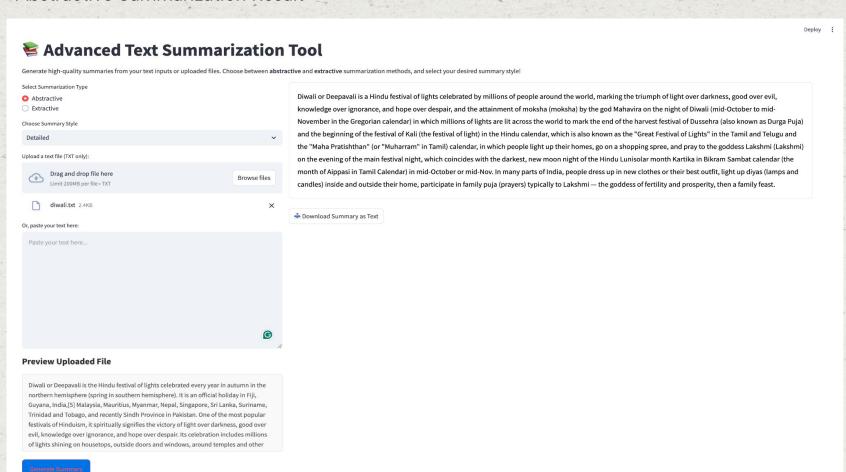
3. ROUGE-L

- Measures overlap based on the Longest Common Subsequence (LCS) between the generated and reference summaries.
- Significance: Captures the order and sequence of words, emphasizing structure and coherence.





Abstractive Summarization Result



Extractive Summarization Result

Advanced Text Summarization Tool Generate high-quality summaries from your text inputs or uploaded files. Choose between abstractive and extractive summarization methods, and select your desired summary style! Select Summarization Type Abstractive Extractive Choose Summary Style Detailed Upload a text file (TXT only): Drag and drop file here Browse files Limit 200MB per file • TXT diwali.txt 2.4KB Download Summary as Text Or, paste your text here: **Preview Uploaded File**

Diwali or Deepavali is the Hindu festival of lights celebrated every year in autumn in the northern hemisphere (spring in southern hemisphere). It is an official holiday in Fiji, Guyana, India,[5] Malaysia, Mauritius, Myanmar, Nepal, Singapore, Sri Lanka, Suriname, Trinidad and Tobago, and recently Sindh Province in Pakistan. Its celebration includes millions of lights shining on housetops, outside doors and windows, around temples and other buildings in the communities and countries where it is observed. In many parts of India, the festivities start with Dhanteras (in Northern and Western part of India), followed by Naraka Chaturdasi on second day, Deepavali on the third day, Diwali Padva dedicated to wife-husband relationship on the fourth day, and festivities end with Bhai Doojdedicated to sister-brother bond on the fifth day. One of the most popular festivals of Hinduism, it spiritually signifies the victory of light over darkness, good over evil, knowledge over ignorance, and hope over despair

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Challenges

1. Data Challenges

- Data Diversity: Ensuring the dataset is representative of various domains and text styles.
- **Preprocessing Issues:** Handling noisy and inconsistent data formats.

2. Model Challenges

- Long Texts: Summarizing lengthy documents without losing context or coherence.
- Balancing Trade-Offs: Striking a balance between informativeness and brevity in summaries.
- **Abstractive Complexity:** Generating human-like summaries while maintaining factual accuracy.

3. Computational Challenges

- Resource Constraints: High memory and computational requirements for training transformer models.
- **Fine-Tuning:** Optimizing hyperparameters for better performance across diverse datasets.

Key Learnings

1. Technical Insights

- Understanding Summarization: Gained a deep understanding of extractive and abstractive methods.
- Model Familiarity: Learned about the advanced transformer architectures like BART and BERT.
- **Importance of Preprocessing:** Learned how clean and well-prepared data directly impacts model performance.

2. Practical Takeaways

- **Evaluation Metrics:** Realized the importance of using ROUGE for consistent performance evaluation.
- Fine-Tuning Skills: Improved ability to fine-tune large models effectively for specific tasks.
- **Challenges in Summarization:** Recognized the complexities of generating context-aware, accurate summaries.

Future Scope

1. Multilingual Summarization

• Extend support for multiple languages to serve a global audience.

2. Domain-Specific Models

• Build customized summarization tools for specialized domains like legal, medical, or technical content.

3. Real-Time Summarization

• Enable summarization of streaming data, such as live events or social media content.

Conclusion

- Successfully developed a summarization tool supporting both extractive and abstractive methods.
- Evaluated performance using ROUGE metrics, demonstrating high-quality summaries.
- Addressed challenges in preprocessing, model training, and evaluation.
- Demonstrated potential applications across domains like media, education, and customer service.

THANK YOU