Capstone Project Proposal for Content Summarization Use Large Languages Model

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1. Definition

Project Overview

This project focuses on report summarization using a Language Model (LLM) to generate concise summaries of multiple news. The goal is to provide user with an efficient way to overview the main content of daily news on internets. In the domain of natural language processing (NLP), news summarization is a crucial application. It involves condensing the main points, findings, and insights from a given set of reports into concise summaries.

This technology enables businesses and organizations to handle vast amounts of information efficiently and make informed decisions based on summarized content. Various techniques have been employed in report summarization, including extractive methods that select and combine existing sentences from the original reports, and abstractive methods that generate new sentences to convey the essence of the reports. Machine learning models, particularly Language Models (LMs), play a significant role in achieving both extractive and abstractive summarization.

Language Models like GPT (Generative Pre-trained Transformer) have shown remarkable capabilities in understanding and generating human-like text. They are pre-trained on extensive datasets and fine-tuned for specific tasks, making them a suitable choice for report summarization.

The Problem Statement

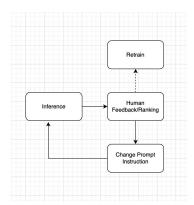
This project tackles the challenge of information overload by focusing on report summarization in Vietnamese using advanced Language Models (LMs). The goal is to create a system that efficiently generates concise summaries for multiple news articles, providing users with a streamlined overview of the main content. In the domain of Natural Language Processing (NLP), news summarization is critical for distilling essential points from reports. The project leverages techniques, including extractive and abstractive methods, with a particular emphasis on the role of Language Models like GPT. These models, pre-trained and fine-tuned, demonstrate remarkable capabilities for summarization, promising to enhance the efficiency of news consumption and decision-making.

Metrics

When evaluating the effectiveness of a summary, various metrics can be employed to measure the quality of the generated output. Here are some commonly used metrics for evaluating summary problems:

Human Judgments/Rankings: Obtain human judgments by having individuals rank multiple summaries based on their perceived quality. This can be done using pairwise

comparisons, where annotators compare two summaries and select the better one, or by directly ranking a set of summaries.



2. Analysis

Data Exploration

The dataset encompasses news data from various sources in Vietnam, obtained through web crawling. Each data point includes the following attributes:

Title: The title represents the headline or name of the news article, providing a quick overview of the content.

Source: The source indicates the origin of the news, specifying the website or platform from which the data was crawled.

Link: The link is the URL leading directly to the original news article, facilitating access to the complete information.

Public: The "Public" attribute denotes whether the news article is publicly accessible. This information is crucial for understanding data visibility.

Author: The author attribute identifies the individual or entity responsible for creating the news article, offering insights into the information's origin.

Summary: The summary provides a concise overview of the key points summaried by human.

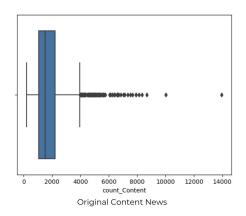
Content: The content attribute encapsulates the full body of information within the news article, comprising text.

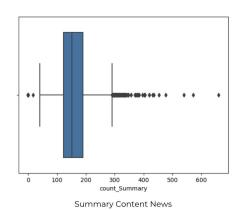
This below is an example dataset:

```
Title: Các nước chia buôn vụ tai nạn mấy bay của Ai Cập
Source: VOV
Link: http://vov.vn/thegioi/cac-nuoc-chia-buon-vu-tai-nan-may-bay-cua-ai-cap-512420.vov
Published Date: Thứ 6, 11:03, 20/05/2016
Author: Diệu Hương/VOV-Trung tâm Tin
Tags: Ai Cập, hộp đen, tai nạn máy bay, mấy bay rơi, tai nan may bay, hop den.tìm kiểm, Ai Cap
Summary: Anh, Pháp, Hy Lạp, NATO đều đề nghị giúp đỡ Ai Cập tìm kiểm hộp đen và diễu tra nguyên nhân của vụ tai nạn mây bay thảm kl
Content:
Tổng thư ký Tổ chức Hiệp ước Bắc Đại Tây Dương (NATO) Jens Stoltenberg ngày 19/5 cho biết, nêu Ai Cập đề nghị, liên minh này sẽ hô
đatasưs khi lời chia buôn sâu sắc nhất đến những ai bị ảnh hưởng bởi vụ việc này. Tổi cũng gửi lời chia buôn sâu sắc đến Pháp và Ai Cật
Thước đó, Hãng hàng không quốc gia Ai Cập (EgyptAir) xác nhận phía Hy Lạp đã tìm thây mảnh vỡ từ chiếc mấy bay này ở phía Nam đảo l
Người dứng đầu cơ quan điều tra tai nạn hàng không Ai Cập Ayman al-Moqadem ngày 19/5 cho biết, nước này sẽ dân đầu một ủy ban điều
Ủy ban này bao gồm cả nhân sự phía Pháp, nước sản xuất chiếc Airbus 320 này và cũng là nước có số nạn nhân nhiều thứ hai sau Ai Cập
Hội đồng an toàn giao thông quốc gia Mỹ cho biết, động cơ của chiếc máy bay gặp nạn được sản xuất tại nước này. Theo quy tắc quốc t
Lúc này, ứng viên Tổng thông đồng Cộng hòa Mỹ Donald Trump đã lên tiếng bày tỏ nghi ngờ đây là một vụ tân công khủng bốo song chính
Thủ tướng Ai Cập Mohamed Morsi đã yêu cầu Bộ Hàng không dân dụng và quân đội phối hợp nhanh chống định vị nơi chiếc mấy bay mang s
Trong khi đó, Ngoại trưởng Canada Stephane Dion ngày 19/5 cho biết trong số những hành khách đi chuyên bay mang số hiệu MS 804 của
Trước đó, hãng hàng không quốc gia Ai Cập đã công bố quốc tịch của những hành khách đi trên chuyên bay mang số hiệu MS 804 của
```

Exploratory Visualization

Character Length Plot





- The average length original content ~ 3000 characters and summary content ~ 300 characters

Algorithms and Techniques

In this section, we present our solution which tries to integrate the pros of previous solutions In general, the steps we did can be listed as follows:

- 1. Use transfer learning to fine-tune a pre-trained Llama model on our dataset.
- 2. Hyperparameter tuning
- 3. Training and evaluation

Benchmark

Method: Finetuning a Pre-Trained Llama

3. Methodology

In this section, we introduce the experiment details including the data preprocessing, implementation procedure, improvements and refinements.

Data Preprocessing

In this data preprocessing step, we specifically extract two attributes: "Summary" and "Content." Subsequently, we apply a filtering process to eliminate outliers, including instances with excessively long lengths or empty values.

Implementation

The benchmark model is implemented as shown in the following snippet code:

```
trainer = transformers.Trainer(

model=model,
train_dataset=tokenized_train_dataset,
eval_dataset=tokenized_test_dataset,
args=transformers.TrainingArguments(
    output_dir=f"./BASE_MODEL_NAME}/{PROJECT_ID}",
    warmup_steps=50,
    per_device_train_batch_size=2,
    gradient_accumulation_steps=4,
    max_steps=1000,
    learning_rate=2.5e=5, # Want about 10x smaller than the Mistral learning rate
    logging_steps=50,
    bf16=false,
    optim='paged_adamw_aBbit',
    logging_dir="./logs", # Directory for storing logs
    save_strategy="steps", # Save the model checkpoint every logging step
    save_steps=50, # Save checkpoints every 50 steps
    evaluation_strategy="steps", # Evaluate the model every logging step
    eval_steps=5, # Evaluate and save checkpoints every 50 steps
    do_eval=True, # Perform evaluation at the end of training
    report_to="wandb", # Comment this out if you don't want to use weights 8 baises
    run_name=f"{BASE_MODEL_NAME}-{PROJECT_ID}-{datetime.now().strftime('%Y-%m-%d-%H-%M')}" # Name of the W&B run (opti
    data_collator=transformers.DataCollatorForLanguageModeling(tokenizer, mlm=False),
}

model.config.use_cache = False # silence the warnings. Please re-enable for inference!
trainer.train()
```

We have also integrated the option to utilize a debugger and profiler, providing a comprehensive examination of computations on our machines.

These implementations are extended to our enhanced model, which will be discussed in more detail later.

4. Result

Model Evaluation and Validation



5. Conclusion

In the course of this project, we undertook exploratory data analysis (EDA), carefully selected the most fitting pre-trained model, and delved into inventive strategies for

enhancement. Our focus was on refining the model specifically for the Vietnamese language, with an emphasis on the summary task.

Notably, we implemented Large Language Models (LLM) for the summary task, harnessing the power of extensive contextual information to achieve improved results.