

# Extracting Balanced Summaries from Politically Biased News using CNN and Pegasus Transformer

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## INTRODUCTION

Extracting the "truth" from the news can be a challenging endeavor. With numerous news outlets and information sources, cross-referencing information is time-consuming and challenging.

This project aims to leverage classification and transformers to generate unbiased summaries of articles across the political spectrum.

**Data Collection:** Used web scraping techniques to obtain our data.

**CNN for Classification Model:** Built a CNN model to classify articles based on political bias.

**Pegasus Transformer:** Utilized the PEGASUS model for abstractive summarization, which masks important sentences and generates a cohesive summary.

## GOALS

- **Develop a classification model** to identify the political bias of news articles.
- **Use Pegasus transformer** to generate unbiased summaries of these articles.
- **Evaluate the effectiveness** of the generated summaries in reducing bias.

## PROJECT SUMMARY

Our project involves creating a system that classifies news articles based on political bias and then generates summaries that are intended to be unbiased.

We faced **challenges** such as limited data sources, summary evaluation, and modifying our classification model for better accuracy.

## METHODS

### Data Collection

We webscraped 300 articles: 100 each from left, center, and right-biased news sources.

**Tools used:** Requests, BeautifulSoup for scraping; CSV for data storage.

## METHODS

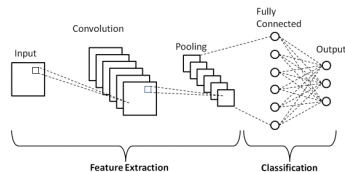
### CNN for Classification Model

#### Convolutional Neural Network (CNN):

**Purpose:** Classify news articles into left, center, and right political biases.

#### Architecture:

- **Input Layer:** Processes the text data.
- **Embedding Layer:** Converts text into numerical vectors using embeddings.
- **Convolutional Layers:** Applies filters to capture essential features and patterns in the text.
- **Pooling Layers:** Reduces the dimensionality of the data, preserving important features.
- **Fully Connected Layers:** Integrates the features for final classification.
- **Output Layer:** Provides the probability distribution over the three classes (left, center, right).



## METHODS

### Pegasus Transformer

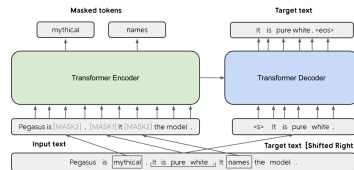
**Purpose:** Generate unbiased summaries of news articles, condensing them while maintaining the core message and minimizing bias.

**Abstractive Summarization:** Creates new sentences to summarize the main ideas, rather than copying parts of the text, producing more natural and coherent summaries.

**Masked Sentence Generation:** Masks important sentences during training and learns to predict them using the rest of the text, helping the model focus on key information for accurate summaries.

#### Encoder-Decoder Framework:

- **Encoder:** Converts input text into hidden states, capturing the text's meaning.
- **Decoder:** Generates the summary from these hidden states.



## EXPERIMENT SETUP

**Data Preparation:** Collected and labeled 300 articles for bias.

**CNN for Classification Model:** Trained the CNN model on the dataset to classify articles into left, center, and right bias.

**Summarization Model:** Used the transformer PEGASUS model to generate summaries of the classified articles.

**Evaluation:** Calculated accuracy for CNN model and used BERTScore metric for summarization evaluation.

## EVALUATION AND RESULTS

After training our CNN model with the dataset, we fed 3 different articles into our CNN model to classify them. After doing so, we used the Pegasus model to generate the following summary:



Tested accuracy for our CNN model: 0.7

BERTScore for generated summary vs input article:

BERTScore For Generated Summary vs. Input Articles: 0.9138839/647/8508

## NEXT STEPS

- **Improve Classification Model:** Enhance the CNN model to better handle nuanced bias and include more in-between affiliations.
- **Advanced Evaluation Techniques:** Implement more sophisticated methods to evaluate the neutrality of the summaries.
- **Expand Data Sources:** Gather more data from a broader range of news outlets to improve model training and evaluation.

## REFERENCES

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