



# **Summarized Context Aware News**

**There are two kinds of  
news consumers.**

wake up sid?



**wake up sid?**



**Woke Sid.**



wake up sid?



?



Woke Sid.



## Lottery firm tops electoral bond list as Election Commission publishes data

The startling fact was that the biggest purchaser of the electoral bonds turned out to be an unknown Coimbatore-based lottery firm that splashed out Rs 1,368 crore in 21 separate bouts of purchases between October 21, 2020, and January 9, 2024

Vivek Nair, Sambit Saha, Pinak Ghosh | Calcutta, Mumbai | Published 15.05.24, 05:50 AM

and obscure firms elbowed out the Big Boys of Business as they mounted a big purchase of electoral bonds from the State Bank of India over a period of five years, which they then funnelled to political parties in the country.

Thursday, the veil of anonymity was finally ripped from large corporate firms and fat cats of the country as the Election Commission of India put out a list of information on the controversial electoral bonds, complying with the order of the Supreme Court that determined the scheme was

POLITICS

## 'Unique Numbers Recorded by SEBI Bought Electoral Bond

"SBI in an RTI reply confirmed that these unique numbers were recorded

## 'Fear return of black money': Amit Shah after Supreme Court scraps electoral bonds

Union Home Minister Amit Shah, in his first reaction after the Supreme Court scrapped the electoral bonds scheme, said it was introduced to "wipe out black money".

Listen to Story



### Polarization of news



Union Home Minister Amit Shah said the electoral bonds scheme was introduced to wipe out black money

India Today News Desk

New Delhi, UPDATED: Mar 15, 2024 23:28 IST

chronology  
kya hai?

**Lack of context**



How can we combat **bias** and **misrepresentation** of news coupled with lack of **awareness**, to make information accessible?



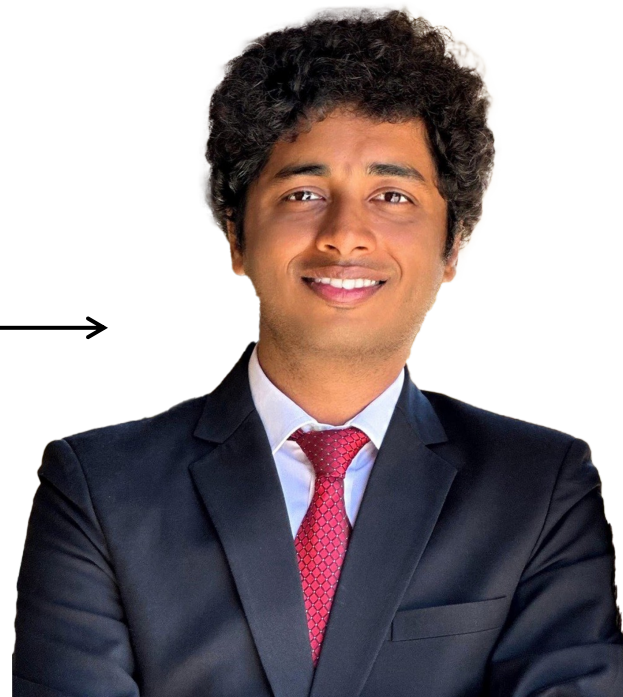
wake up sid?



scan.



Woke Sid.





## Potential Applications

- Can be used in collaboration with platforms like Google News & InShorts
- Implemented as a Browser extension for ease of use
- Media Bias and Objectivity Detection

## Potential Impact

- Regulation of media and news consumption
- Informed decision-making
- Increase in number of news readers

# Literature Review

# Sentiment Analysis

- Comparison of the classifier models CNN and NBC for sentiment analysis on English-language tweets concerning the Turkey Crisis of 2018.
- CNN outperformed NBC on various performance metrics.
- CNN: A Deep Learning algorithm.
- NBC: A simple probabilistic classifier based on the Bayes' theorem.

## Comparison of Accuracy between Convolutional Neural Networks and Naïve Bayes Classifiers in Sentiment Analysis on Twitter

P.O. Abas Sunarya<sup>1</sup>, Rina Refianti<sup>2</sup>, Achmad Benny Mutiara<sup>3</sup>, Wiranti Octaviani<sup>4</sup>  
Dept. of Informatics Engineering, STMIK Raharja Jl. Jenderal Sudirman No. 40, Tangerang 15117, Indonesia<sup>1</sup>  
Faculty of Computer Science and Information Technology  
Gunadarma University, Jl. Margonda Raya No. 100, Depok 16424, Indonesia<sup>2,3,4</sup>

TABLE II. COMPARISON OF CNN AND NBC CLASSIFICATION REPORTS AGAINST DATA TEST

	Precision		Recall		f1-Score		Support	
	CNN	NB	CNN	NB	CNN	NB	CNN	NB
Neutral	0.90	0.84	0.89	0.69	0.89	0.75	629	650
Positive	0.89	0.77	0.87	0.83	0.88	0.80	659	627
Negative	0.85	0.75	0.89	0.82	0.87	0.78	669	680
<b>Total</b>	<b>0.88</b>	<b>0.78</b>	<b>0.88</b>	<b>0.78</b>	<b>0.88</b>	<b>0.78</b>	<b>1957</b>	<b>1957</b>

# Abstractive Summarization

- Feature-rich encoder captures important keywords better using linguistic features
- Switching generator-pointer handles rare/unseen words effectively
- Hierarchical attention captures document structure for long inputs
- Authors primarily used data from CNN and Daily Mail

## Abstractive Text Summarization using Sequence-to-sequence RNNs and Beyond

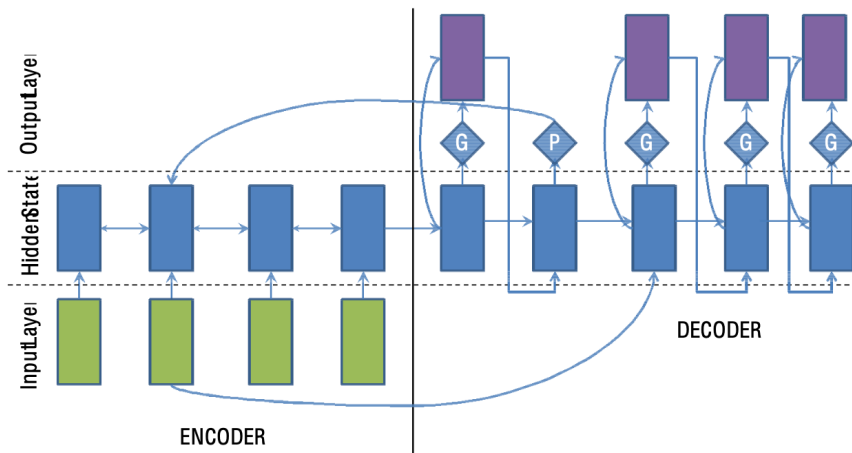
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# Evaluation Metrics

- ROUGE Score – Traditional Scoring method
- Alternative – G-Eval by OpenAI
- Uses LLMs and scoresheet-based prompt – Slightly Biased towards AI generated content
- Survey people at Plaksha to get summary scores and opinions
- Draw comparisons with Human written summaries for better evaluation

## G-EVAL: NLG Evaluation using GPT-4 with Better Human Alignment

Yang Liu Dan Iter Yichong Xu  
Shuohang Wang Ruochen Xu Chenguang Zhu

Microsoft Cognitive Services Research  
{yaliu10, iterdan, yicxu, shuowa, ruox, chezhu}@microsoft.com

Metrics	Naturalness		Coherence		Engagingness		Groundedness		AVG	
	<i>r</i>	$\rho$	<i>r</i>	$\rho$	<i>r</i>	$\rho$	<i>r</i>	$\rho$	<i>r</i>	$\rho$
ROUGE-L	0.176	0.146	0.193	0.203	0.295	0.300	0.310	0.327	0.243	0.244
BLEU-4	0.180	0.175	0.131	0.235	0.232	0.316	0.213	0.310	0.189	0.259
METEOR	0.212	0.191	0.250	0.302	0.367	0.439	0.333	0.391	0.290	0.331
BERTScore	0.226	0.209	0.214	0.233	0.317	0.335	0.291	0.317	0.262	0.273
USR	0.337	0.325	0.416	0.377	0.456	0.465	0.222	0.447	0.358	0.403
UniEval	0.455	0.330	0.602	0.455	0.573	0.430	0.577	0.453	0.552	0.417
G-EVAL-3.5	0.532	0.539	0.519	0.544	<b>0.660</b>	<b>0.691</b>	<b>0.586</b>	0.567	0.574	0.585
G-EVAL-4	<b>0.549</b>	<b>0.565</b>	<b>0.594</b>	<b>0.605</b>	0.627	0.631	0.531	0.551	<b>0.575</b>	<b>0.588</b>

G-Eval Correlation Score Comparison

# Features Preprocessing

# Global News Dataset

- Open Source - Kaggle
- 98,305 Articles – Scraped by authors
- Sentiment Labels – Positive, Negative and Neutral
- Relevant Columns – Title, Description, Article, Title Sentiment
- No Summaries provided

**data.csv** (428.83 MB) ↓ 🔍 ➤

Detail **Compact** Column 10 of 12 columns ▼

**About this file** 📝 Add Suggestion

Same description as raw-data, but now with an extra column called full\_content extracted using [dailyWorker.py](#).

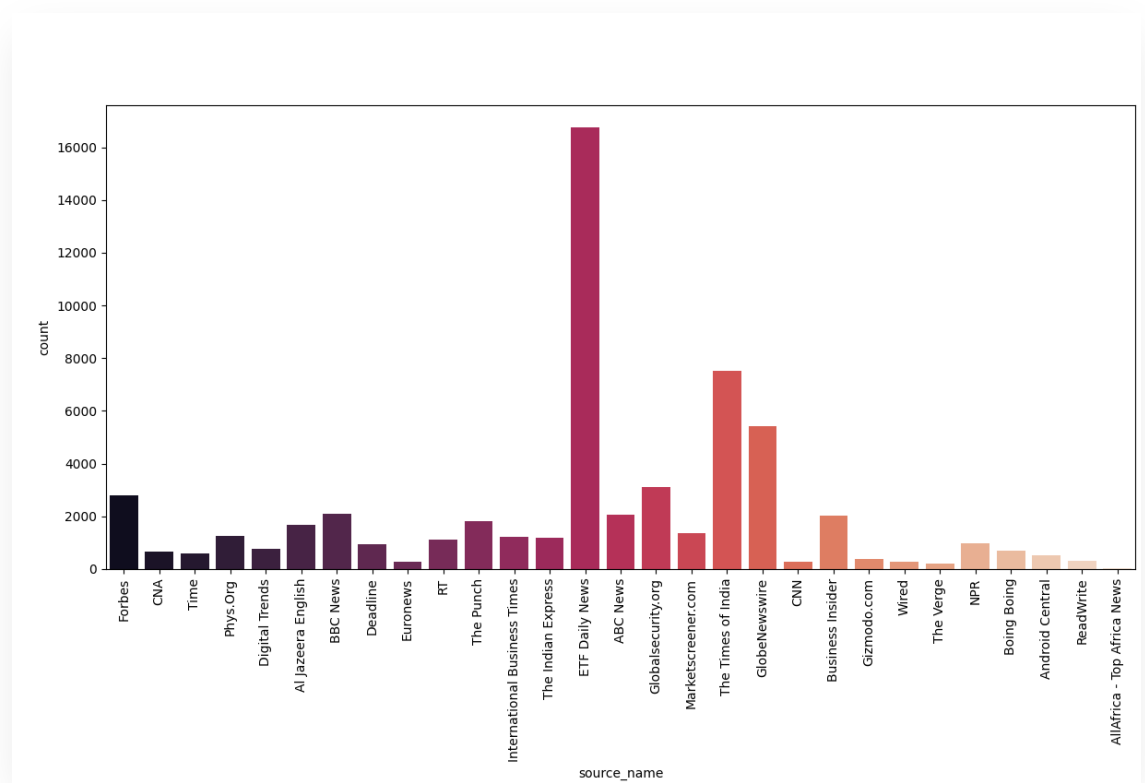
source_name	author	title	description	url
source name	The author of the article	The headline or title of the article.	A description or snippet from the article.	The direct URL to the article.
Daily News 16%	MarketBeat News 16%	<b>98305</b> unique values	<b>98489</b> unique values	<b>101832</b> unique values
Times of India 7%	[null] 8%			
er (81115) 77%	Other (80529) 76%			
ernational iness Times	Paavan MATHEMA	UN Chief Urges World To 'Stop The Madness' Of Climate Change	UN Secretary-General Antonio Guterres urged the world Monday to "stop the	<a href="https://www.ibtimes.com/un-chief-urges-world-stop-madness-climate-change">https://www.ibtimes.com/un-chief-urges-world-stop-madness-climate-change</a>



# News Sources

## Insights

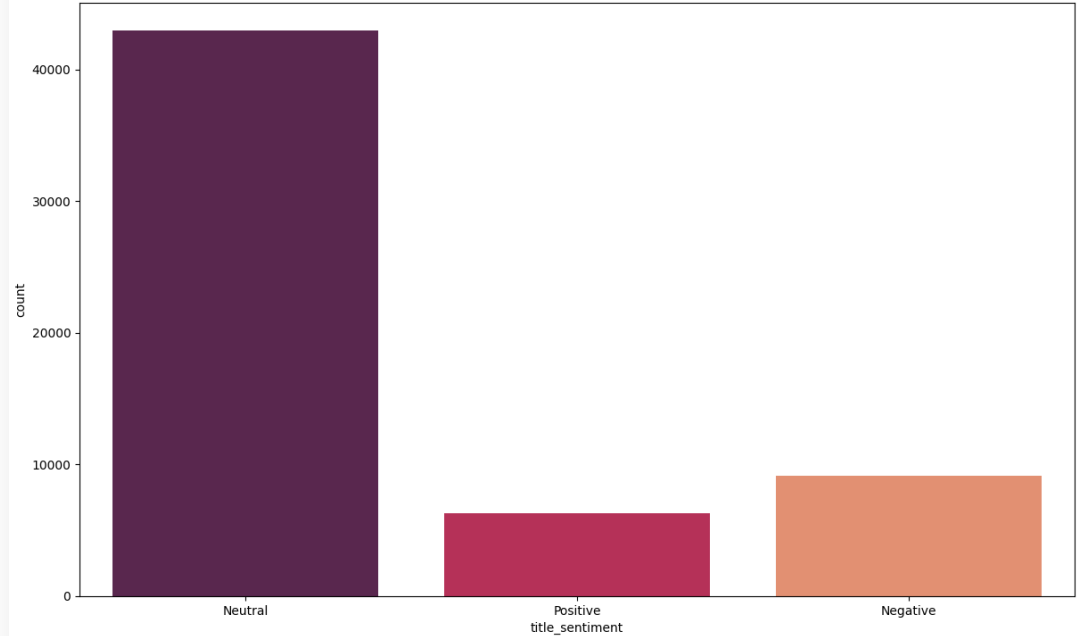
- ETF Daily News - largest news source
- Most articles are financial news
- Indian news outlets – Times of India, Wired, Indian Express etc.



# Title Sentiment

## Insights

- Dataset is heavily biased towards neutral
- ETF Daily News – 90% Neutral  
Digital Trends – 51% Positive

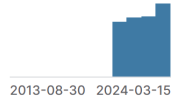


# BBC News Dataset

- Open Source - Kaggle
- ~30,000 Articles – Updated Daily
- 5 Columns in total
- Relevant Columns – Title, link, description
- No Summaries or Sentiment

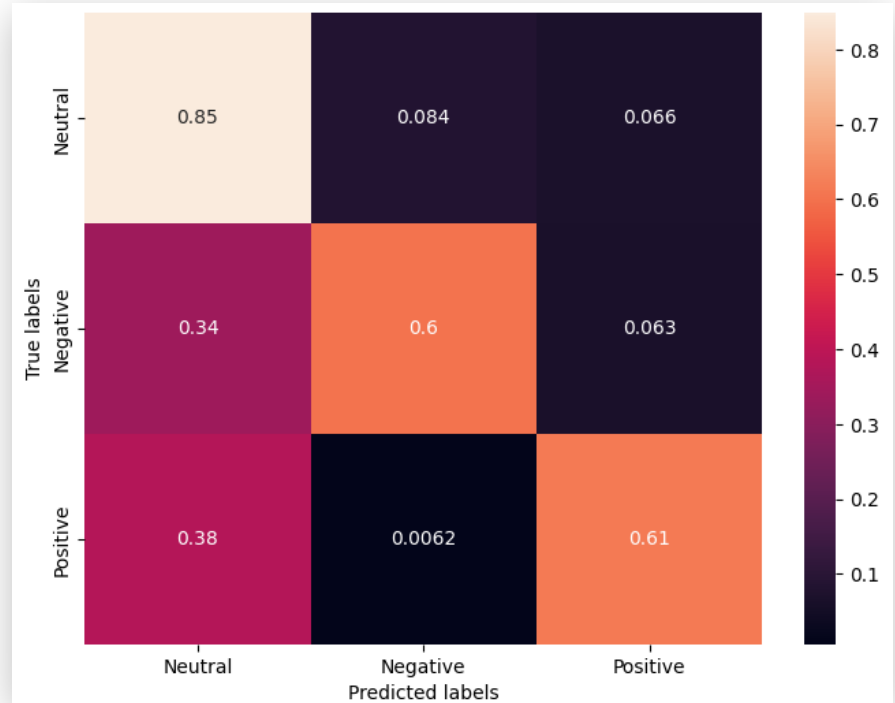
**bbc\_news.csv** (10.19 MB) ↓ ↻ >

Detail Compact Column 5 of 5 columns ▾

<b>title</b> RSS Feed Title	<b>pubDate</b> RSS Feed publication date	<b>guid</b> RSS Feed guid	<b>link</b> RSS Feed link	<b>description</b> RSS Feed description
<b>29456</b> unique values		<b>28155</b> unique values	<b>28111</b> unique values	<b>28834</b> unique values
Ukraine: Angry Zelensky vows to punish Russian atrocities	Mon, 07 Mar 2022 08:01:56 GMT	<a href="https://www.bbc.co.uk/news/world-europe-60638042">https://www.bbc.co.uk/news/world-europe-60638042</a>	<a href="https://www.bbc.co.uk/news/world-europe-60638042?at_medium=RSS&amp;at_campaign=KARANGA">https://www.bbc.co.uk/news/world-europe-60638042?at_medium=RSS&amp;at_campaign=KARANGA</a>	The Ukrainian president says the country will not forgive or forget those who murder its civilians.
War in Ukraine: Taking cover in a town under attack	Sun, 06 Mar 2022 22:49:58 GMT	<a href="https://www.bbc.co.uk/news/world-europe-60641873">https://www.bbc.co.uk/news/world-europe-60641873</a>	<a href="https://www.bbc.co.uk/news/world-europe-60641873?at_medium=RSS&amp;at_campaign=KARANGA">https://www.bbc.co.uk/news/world-europe-60641873?at_medium=RSS&amp;at_campaign=KARANGA</a>	Jeremy Bowen was on the frontline in Irpin, as residents came under Russian fire while trying to flee...
Ukraine war 'catastrophic for	Mon, 07 Mar 2022 00:14:42 GMT	<a href="https://www.bbc.co.uk/news/business-">https://www.bbc.co.uk/news/business-</a>	<a href="https://www.bbc.co.uk/news/business-">https://www.bbc.co.uk/news/business-</a>	One of the world's biggest fertiliser

# Labelling unlabeled data

- Self Training Classifier with **Random Forest** as Base Model
- 58K labeled samples → 80-20 split
- 78K unlabeled samples
- Accuracy Score → **~73%**
- Not using for further training.




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## Russia sanctions: What impact have they had on its oil and gas exports?

26 January 2023  
By Jake Horton & Daniele Palumbo, BBC Reality Check



Since Russia's invasion of Ukraine, many countries have pledged to end or restrict their oil and gas imports to curtail Moscow's revenues and hinder its war effort.

The Russian economy is highly dependent on its energy sector, and Europe is a major importer of Russian energy, making cutting back difficult.

### What sanctions are in place?

EU nations have [ended imports of Russian oil brought in by sea](#), and a ban on refined oil products will come in from 5 February.

The United States said last March it would stop importing Russian oil, and the UK said it was banning Russian crude oil and refined products [with effect from 5 December](#).

# Web Scraping

What our scraper does:

- Fetches article content using NewsAPI endpoint with the provided URL.
- Scrapes article content from the webpage using BeautifulSoup with specified HTML tags.
- Reads CSV file with news article URLs into a pandas data frame.
- Iterates through the data frame rows, extracts article content with BeautifulSoup, and updates the data frame.

# Gemini Summarization

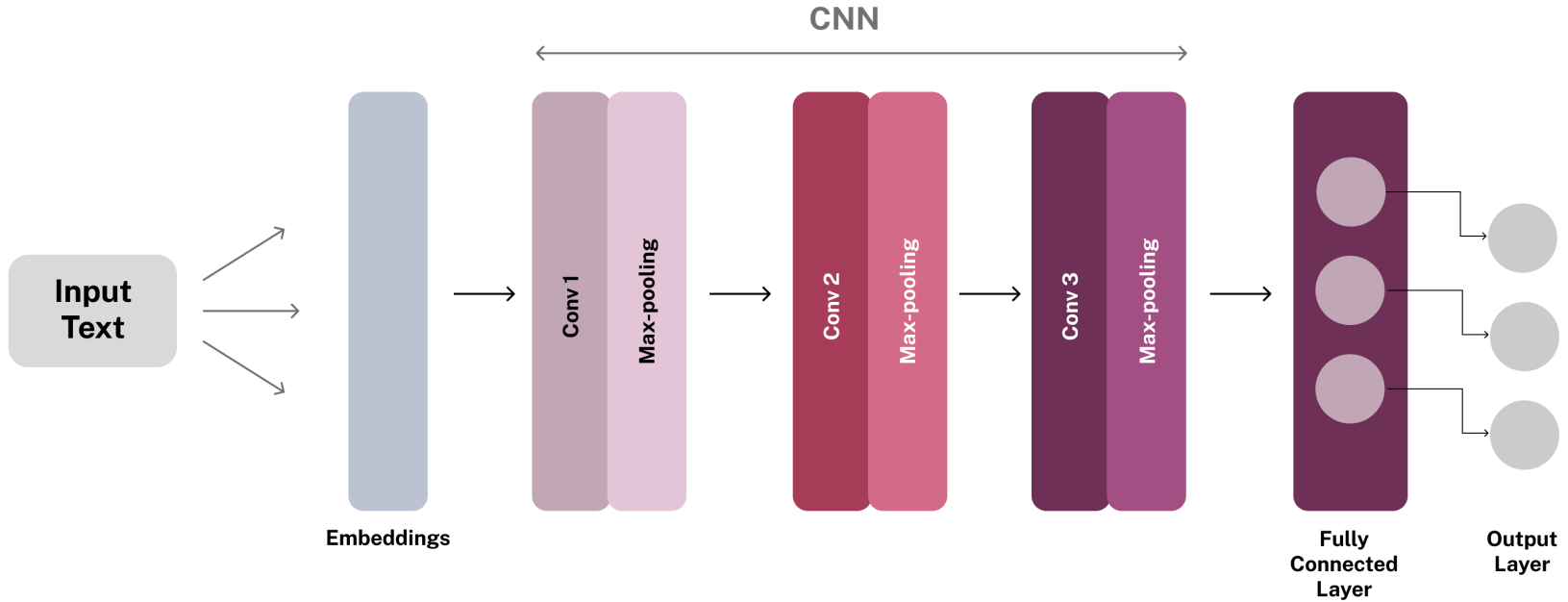
## Advantages

- Human-like linguistic capabilities for summarization
- Processing rate - 12 articles per minute
- Temperature – 0
- Top\_k – 1
- Top\_p – 1
- Max Output – 400 tokens



# **ML Methodology**

# Sentiment Analysis Model Architecture





# Model Configuration

- 3 Convolutional Layers
- Kernel Sizes - 3, 4, and 5 (for each conv layer)
- Vocabulary Size - Number of unique tokens in the vocabulary
- Embedding Dimension – 100
- Dropout - 0.5
- Number of Classes - 3 (Positive, Negative, Neutral)
- Number of Channels - 100 (output channels in each conv layer)

# Model Outputs

## CINEMA

Way Ahead of Bollywood: The Exceptional Realism of the Parallel Tamil Cinema Tackling Social Issues

Predicted: Positive, we believe: Positive

## BIOGRAPHY

The Impending Death of Biography

Predicted: Negative, we believe: Negative

## HISTORY

Leadership in Turbulent Times

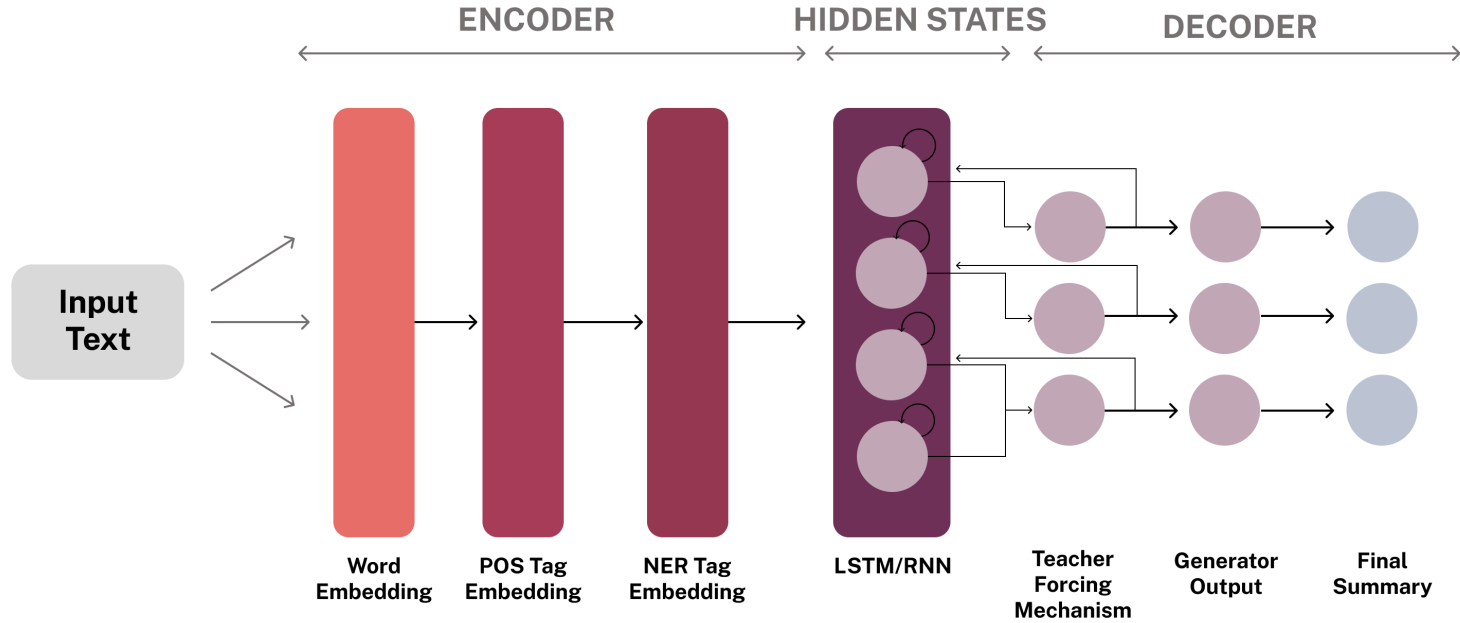
Predicted: Neutral, we believe: Neutral

## CRICKET, BIOGRAPHY

Mahendra Singh Dhoni: Learner, Leader, Legend

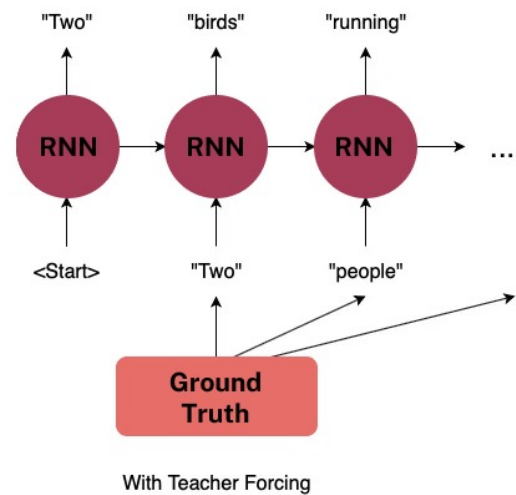
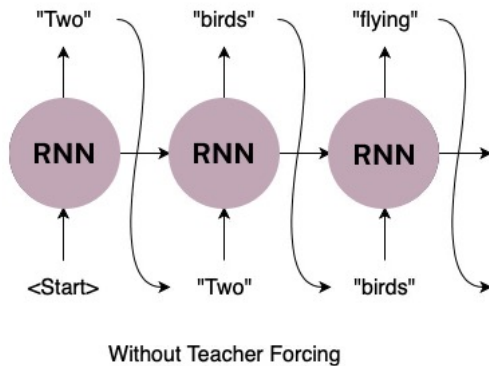
Predicted: Neutral, we believe: Positive

# Summarization Model Architecture



# Model Configuration

- Hidden Layer Dimension – 256
- Embedding Dimension – 300
- Vocab Size – 75,000 words
- Max Summary length – 100 words
- Teacher Forcing Ratio – 0.5
- Dropout – 0.5



## CINEMA

# Way Ahead of Bollywood: The Exceptional Realism of the Parallel Tamil Cinema Tackling Social Issues



Siddharth

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Oct 21, 2022 · 5 mins read

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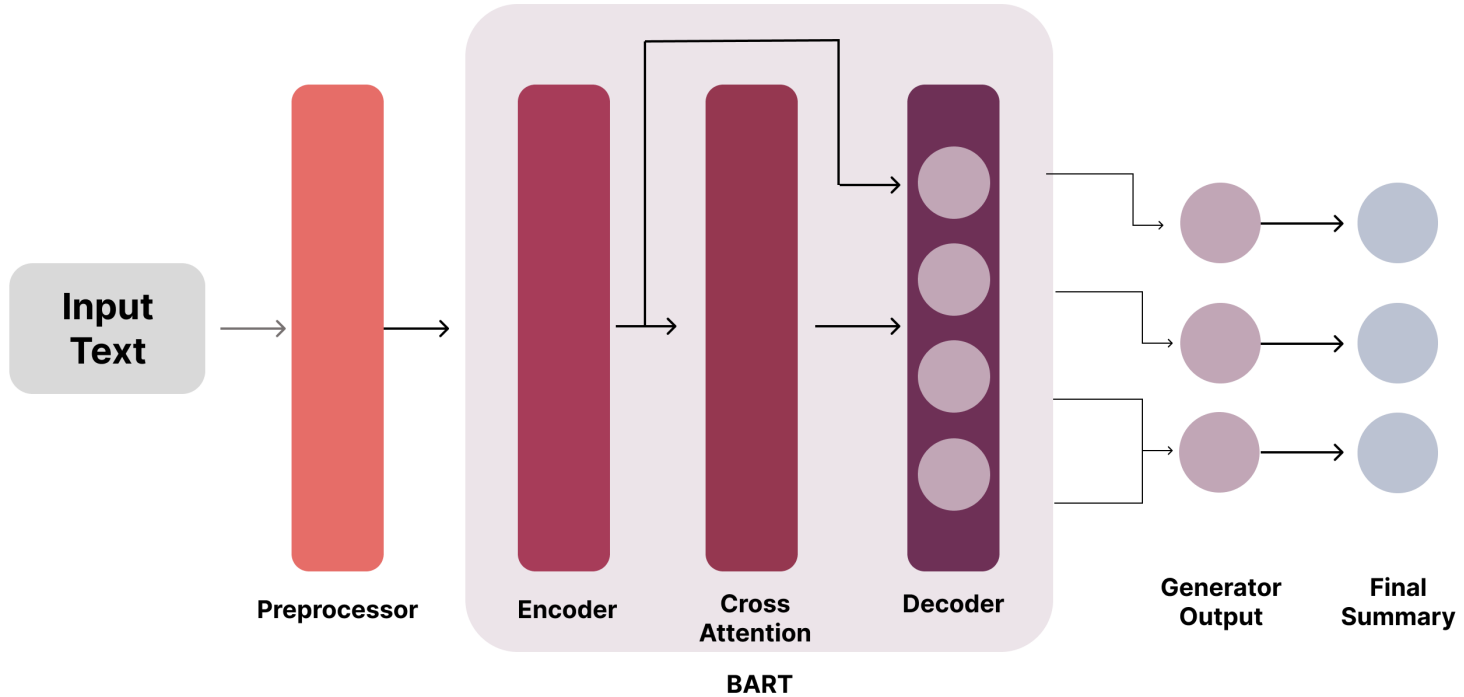
Since last December, four of the five highest-grossing Indian movies have been from South India namely *KGF: Chapter 2*, *RRR*, *Ponniyin Selvan: Part I*, and *Pushpa: The Rise - Part 1*. These hugely entertaining movies have enthralled audiences not just in the South but rather Indians across the world. The ripple effect originating from the success of these movies is not just limited to the profits they have made for their producers or the cultural shift that they have ushered in the visual depiction of Hindu epics

# Example Output

<unk> the ag ceasefire on been to the kong on been to a for the sell of the south india film india the the the the the of world  
audience of the effect the

**okay, toh ab?**

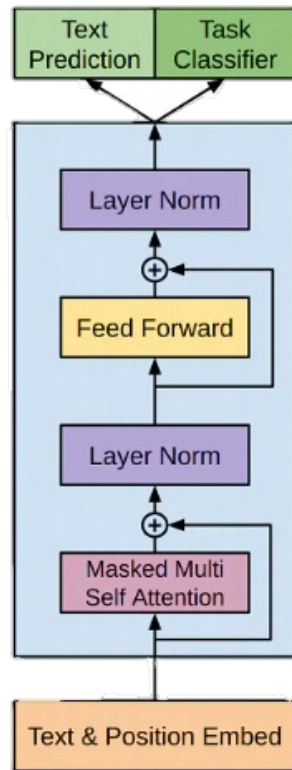
# Summarization Model Architecture





# Model Configuration

- Frozen Encoder Weights
- Max Encoding Sequence Length - 512
- Max Decoder Sequence Length - 128
- Max Generation Length - 40
- Learning Rate -  $5e-5$
- Gradient Clipping Threshold - 1.0
- Excluded bias, gamma, and beta terms



## Example Output

five of the top performing indian movies are based in the south including kGFs Chapter 2 and rRRs two films have been compared to other southbased bollywood films these movies have challenged the box office dominance and inspired a new appreciation of cinema in the indian context

# Challenges Faced

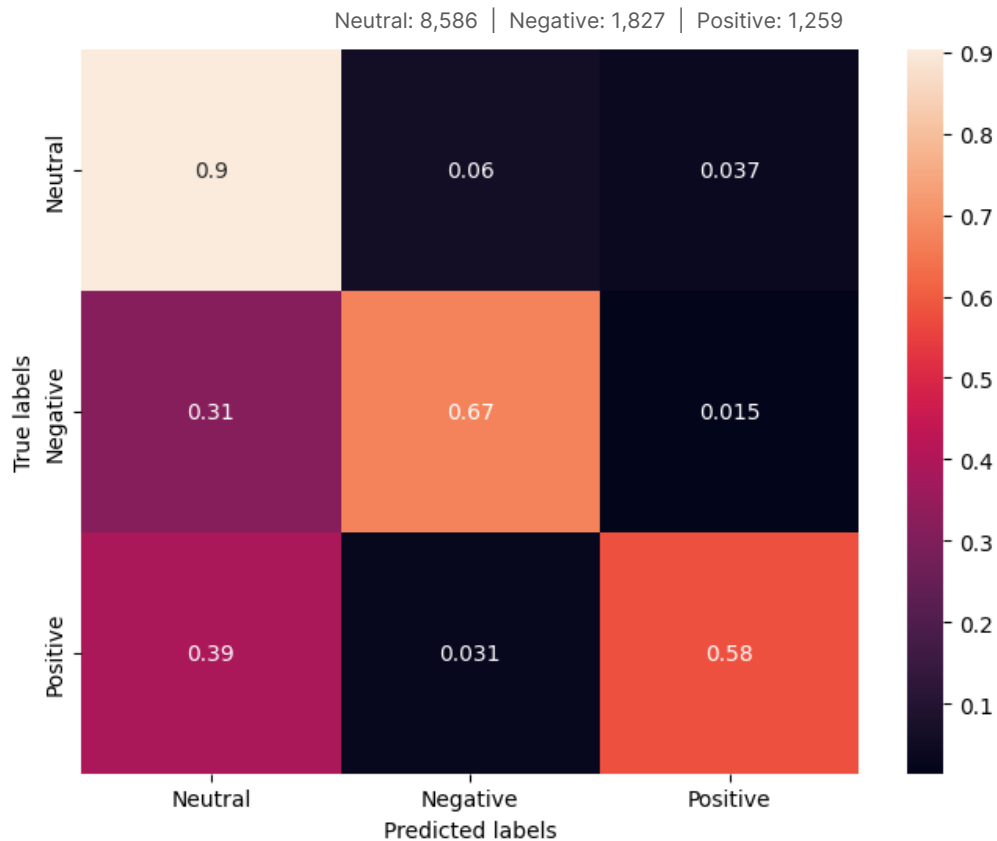
- Models were extremely computationally expensive to train.
- Retaining context over long text sequences proved to be difficult with limited resources
- Tensor Dimensionality issues due to library compatibility.
- Coordinating and integrating the multiple parallel decoder components during training and inference is complex.

# Performance Metrics

# Evaluating Sentiment Classifier

- Trained for 10 Epochs
- Accuracy: 83%
- Precision: 83%
- Recall: 83%
- F1 Score: 83%

^Our model is a Kapil Dev fan



# ROUGE Scores

Recall-Oriented Understudy for Gisting Evaluation

ROUGE was designed to evaluate the quality of machine-generated summaries by comparing them to reference summaries provided by humans.

## ROUGE-N

- Measures overlap of **n-grams** between the candidate and the target text.
- Used to evaluate the grammatical correctness and fluency of generated text.

## ROUGE-L

- Measures the **longest common subsequence** (LCS) between the candidate & reference text.
- Used to evaluate the semantic similarity and content coverage of generated text.

# Recall & Precision

how much of the **reference summary** is the **system summary** capturing?

System Generated:

the cat was found under the bed

Target Summary:

the cat was under the bed

$$\text{recall} = \frac{\text{no. of overlapping words}}{\text{total words in } \mathbf{target} \text{ summary}} = 1.0$$

how much of the **system summary** was in fact **relevant**?

System Generated:

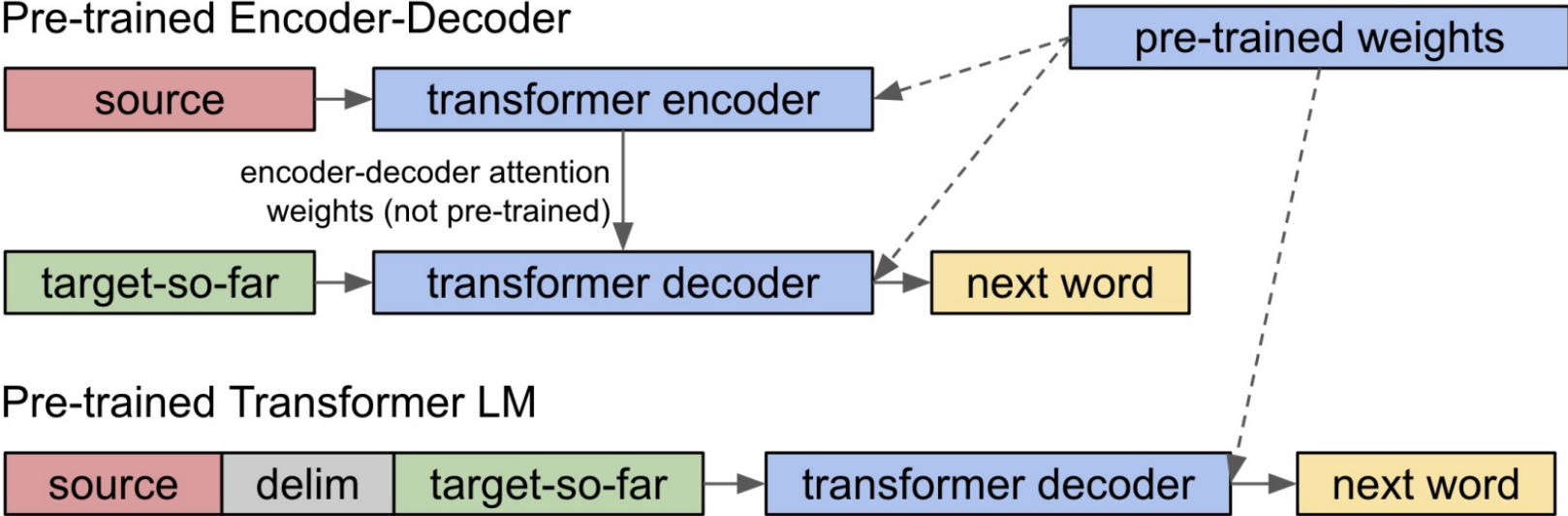
the cat was found under the bed

Target Summary:

the cat was under the bed

$$\text{precision} = \frac{\text{no. of overlapping words}}{\text{total words in } \mathbf{output} \text{ summary}} = 0.85$$

# BERT by Khandelwal et al., 2019





# ROUGE Model Comparison

<b>F1-Score</b>	<b>LSTM - 1</b>	<b>IBM</b>	<b>SCAN</b>	<b>Stanford</b>
<b>ROUGE-1</b>	0.33	0.24	0.48	0.37
<b>ROUGE-2</b>	0.03	0.09	0.23	0.16
<b>ROUGE-L</b>	0.2	0.2	0.38	0.34

# G-Eval Metric

- Coherence - 4/5
- Engagingness – 3/5
- Factual Consistency – 4/5
- Overall – 3.5/5

## ChatGPT

This summary provides a brief overview of the main topic, which is the success of South Indian movies in challenging Bollywood's dominance at the box office and inspiring a new appreciation for cinema in India. However, the coherence could be improved by organizing the information in a clearer and more structured manner. For example, the summary could start by introducing the trend of South Indian movies performing well, then mention specific examples like "KGF Chapter 2" and "RRR," and finally discuss their impact on Indian cinema. Overall, I would rate the coherence of this summary a 3 out of 5.



## ChatGPT

This summary effectively captures the main topic and key points of the news article. It discusses Mohamed Muizzu's victory over incumbent Ibrahim Solih in the Indian Ocean region, Muizzu's pledged projects, and his expressed concerns about India defense partnerships. The summary also highlights the potential shift in Maldives relations with India and the impact of Muizzu's victory on India's strategic interests in the Indian Ocean region. The information is presented in a clear and logical order, making it easy to follow. Therefore, I would rate the coherence of this summary a 5 out of 5.



# Bibliography

- Lin, C. (2004, July 1). ROUGE: a package for automatic evaluation of summaries. ACL Anthology.  
<https://aclanthology.org/W04-1013>
- Loshchilov, I., & Hutter, F. (2017). Decoupled weight decay regularization. arXiv preprint arXiv:1711.05101.
- Khandelwal, U., Clark, K., Jurafsky, D., & Kaiser, L. (2019). Sample efficient text summarization using a single pre-trained transformer. arXiv preprint arXiv:1905.08836.
- <https://www.freecodecamp.org/news/what-is-rouge-and-how-it-works-for-evaluation-of-summaries-e059fb8ac840/>

**thank you.**