

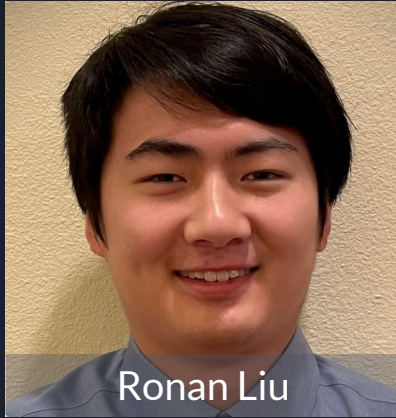
Satire Spotter



satire spotter

Barrett Hooper, Ronan Liu, Emerald Swei

The Satire Spotter Team



Ronan Liu

Software Engineer

Ronan aims to tackle the prevalence of misinformation being spread on social media in recent years.



Emerald Swei

Data Science Consultant

Emerald's primary ambition is to enhance online literacy and transparency, allowing users to make more informed decisions and react with context.

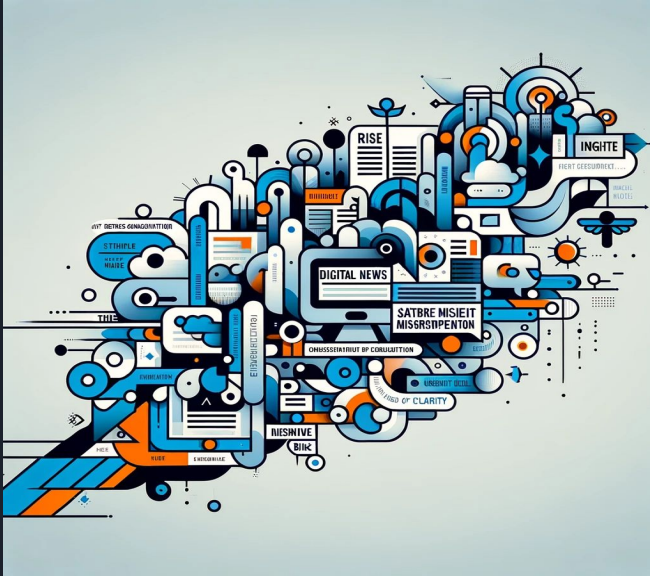


Barrett Hooper

Senior Financial Analyst

Barrett's interest is focused on mitigating the risks of satire misinterpretation, especially in a cross-cultural context.

Problem Space and Impact



Rise of
Digital News
Consumption

Risks of Satire
Misinterpretation

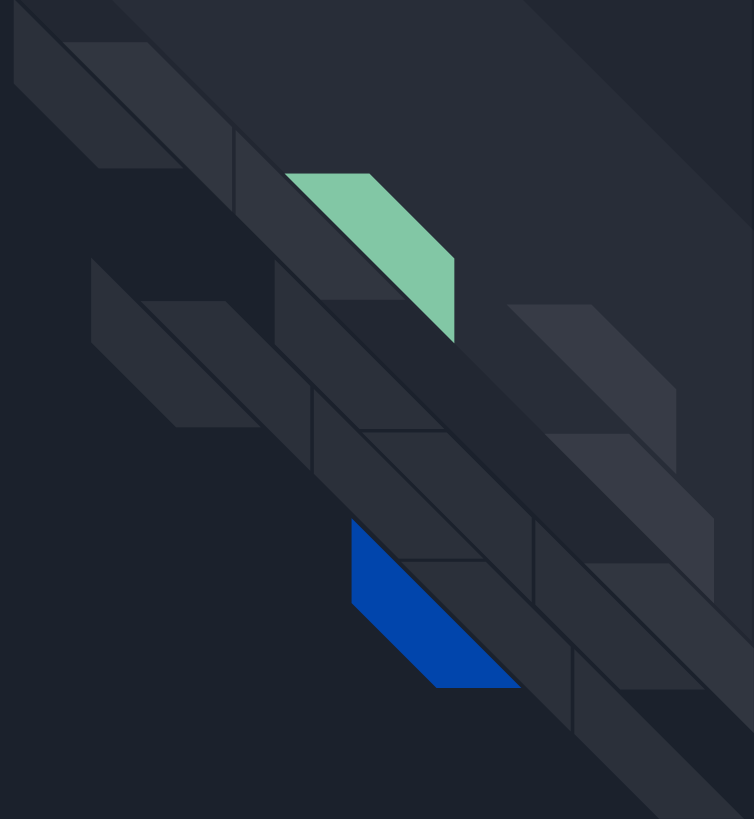
Urgent Need
for Clarity

Proposed Solution

Introduce a label at a granular, post level for all digital content to clarify the intent of each post and reduce confusion.



Technical Approach/Model Evaluation



Datasets Used

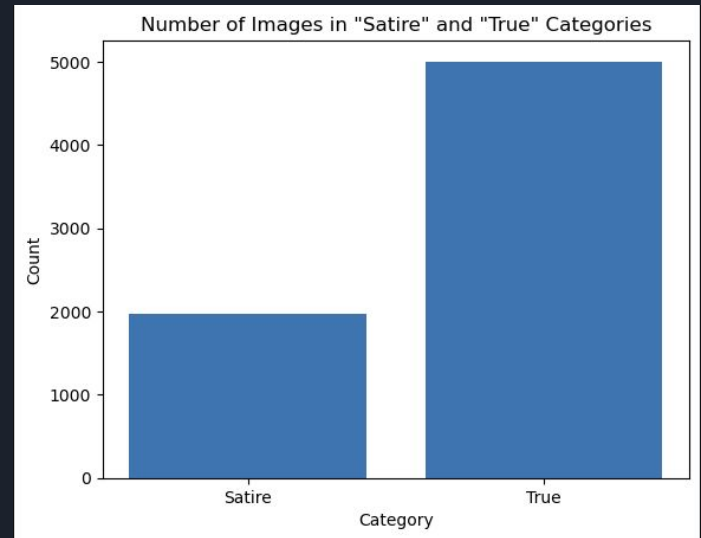
(Approximate Numbers)

Text-only Model:

1. Satire News Detection Corpus
230 Satire / 4,000 True
2. Crawled Current Onion Articles
50 Satire
3. Princeton Senior Thesis Dataset
13,000 Satire
4. Common Crawl News Dataset
20,000 True

Images:

2,000 Satire / 5,000 True



Exploratory Data Analysis: Image Dataset



Figure 1: Examples of satirical news images created by altering existing images.



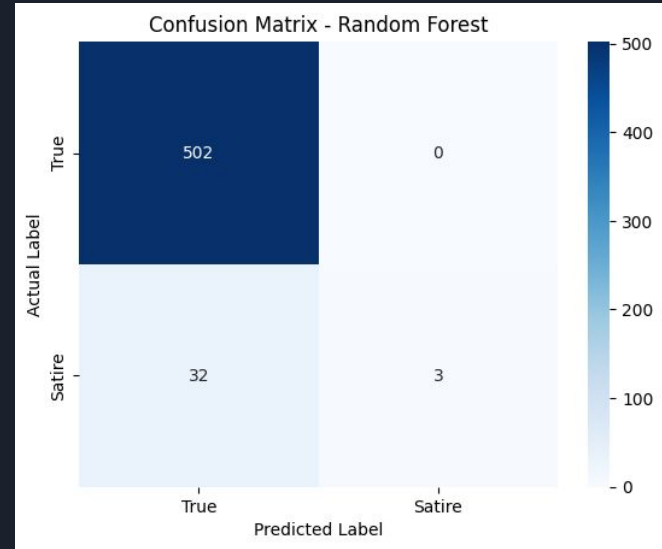
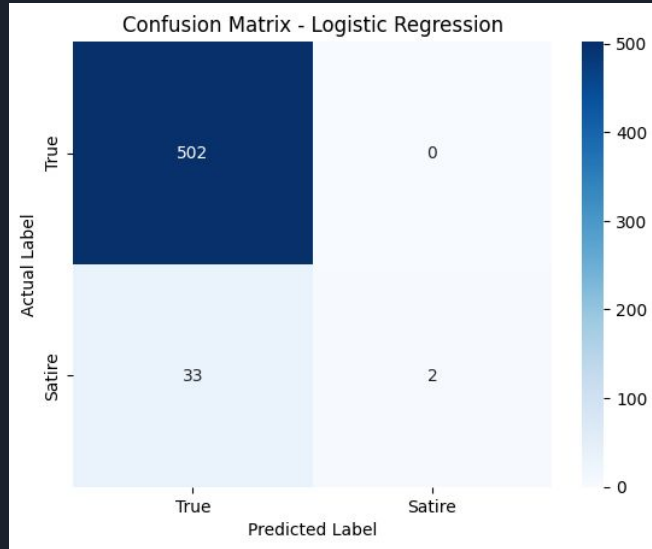
Modeling Approach: Text

Logistic Regression 1.0 &
Random Forest 2.0

Logistic
Regression 2.0
& Random
Forest 2.0

Long Short
Term Memory
(LSTM)

Logistic Regression 1.0 & Random Forest 1.0



Logistic Regression 2.0 & Random Forest 2.0

Hyperparameter Tuning

Regularization

Augmented Dataset

```
Classification Report - Logistic Regression:
precision    recall  f1-score   support

   True      0.93    0.90    0.91     2662
  Satire     0.94    0.96    0.95     4501

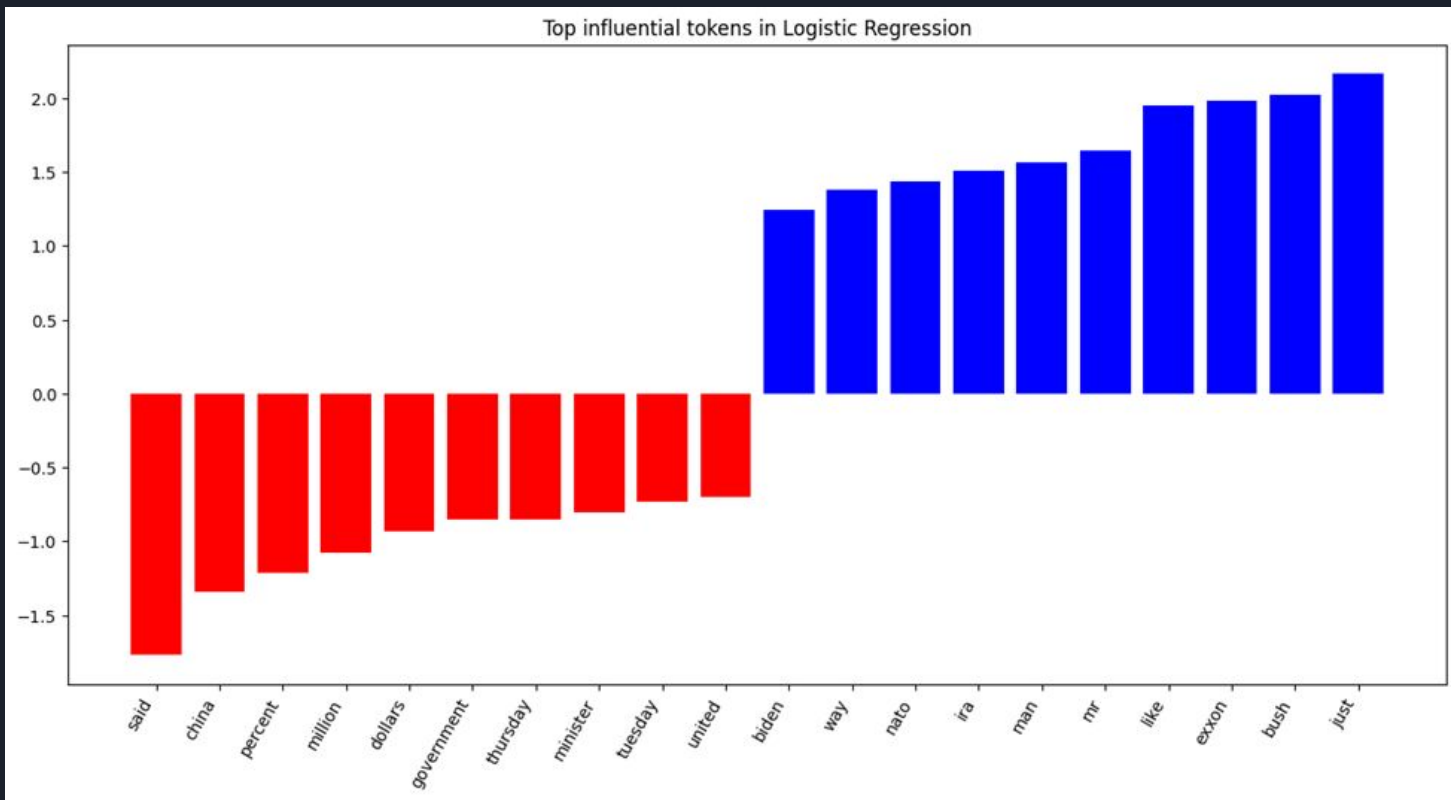
 accuracy    0.94     7163
 macro avg   0.93     7163
weighted avg 0.93     7163

Classification Report - Random Forest:
precision    recall  f1-score   support

   True      0.88    0.84    0.86     2662
  Satire     0.91    0.94    0.92     4501

 accuracy    0.90     7163
 macro avg   0.89     7163
weighted avg 0.90     7163
```

Important Features



LSTM Deep Learning Model

224/224 [=====] - 27s 116ms/step

precision recall f1-score support

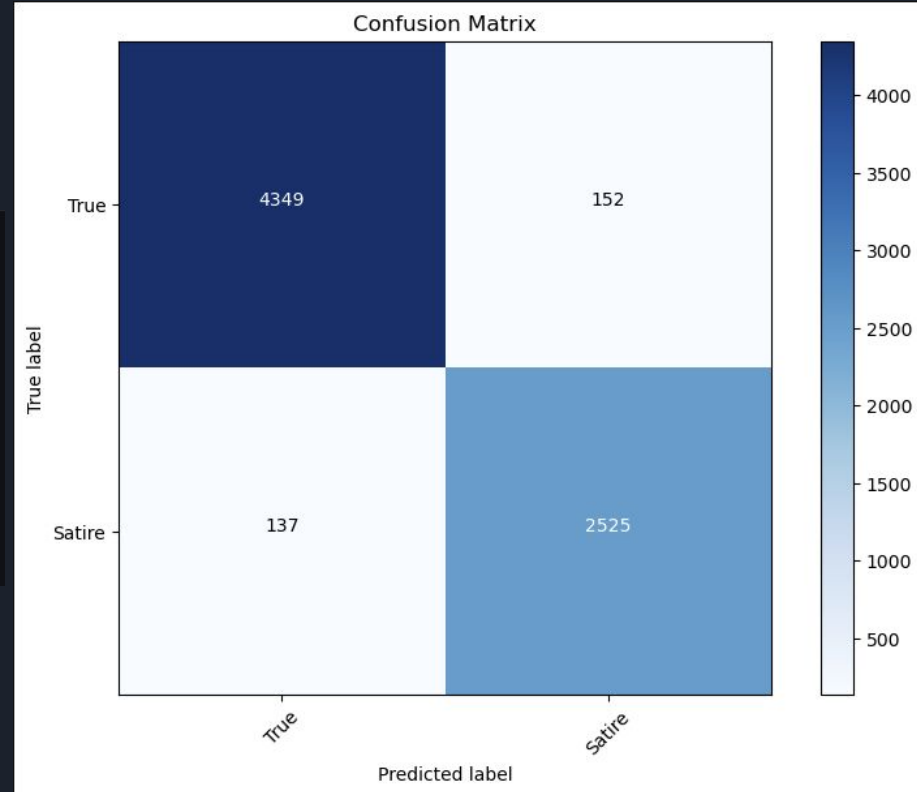
True 0.97 0.97 0.97 4501

Satire 0.94 0.95 0.95 2662

accuracy 0.96 7163

macro avg 0.96 0.96 0.96 7163

weighted avg 0.96 0.96 0.96 7163



Modeling Approach: Images

CNN				
	precision	recall	f1-score	support
True	0.64	0.83	0.72	134
Satire	0.77	0.55	0.64	141
accuracy			0.69	275
macro avg	0.71	0.69	0.68	275
weighted avg	0.71	0.69	0.68	275

SVM				
	precision	recall	f1-score	support
True	0.64	0.66	0.65	134
Satire	0.67	0.65	0.66	141
accuracy			0.65	275
macro avg	0.65	0.65	0.65	275
weighted avg	0.65	0.65	0.65	275

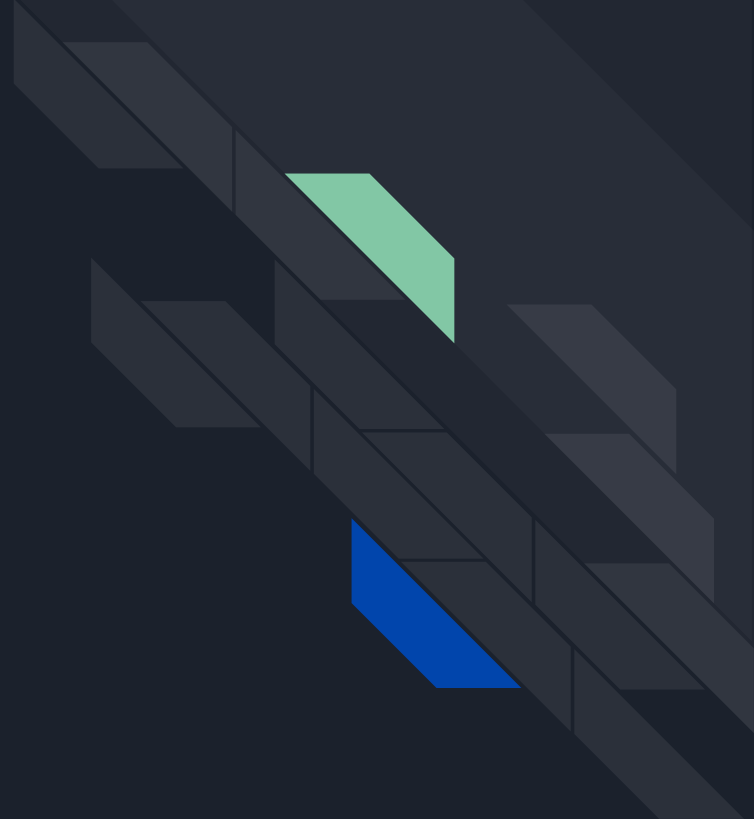
Pre-Trained Image Model

MobileNet V2

Classification Report:

	precision	recall	f1-score	support
0	0.84	0.89	0.86	134
1	0.89	0.84	0.86	141
accuracy			0.86	275
macro avg	0.86	0.86	0.86	275
weighted avg	0.86	0.86	0.86	275

Final Findings



Comparison to Current Models

Text Model

Table II. Accuracy for all

Weighing Parameter	Accuracy
TF-IDF	83.4160 %
BIN	84.0833 %
BNS	89.9167 %
TF-BNS-IDF	92.7500 %

Table III. Comparative study

Weighing Parameter	Precision	Recall	F-score
TF-IDF	0.722	0.732	0.727
BIN	0.746	0.788	0.767
BNS	0.802	0.859	0.829
TF-BNS-IDF	0.826	0.870	0.840

Image Model

Type	Model	Accuracy	F1 score	AUC-ROC
Baselines	All regular news	60.00	—	50.00
	ResNet101	73.54	65.26	80.28
	BERT _{BASE}	91.33	88.64	96.77
	Simple fusion (average)	92.53	90.44	96.74
	Simple fusion (concatenation)	92.74	90.70	97.31
	Proposed Models	ELA+CNN	44.20	51.86
ViLBERT		93.80	92.16	98.03

Table 1: Model performance on satire detection dataset.



How Can Target Users Use Our Model?





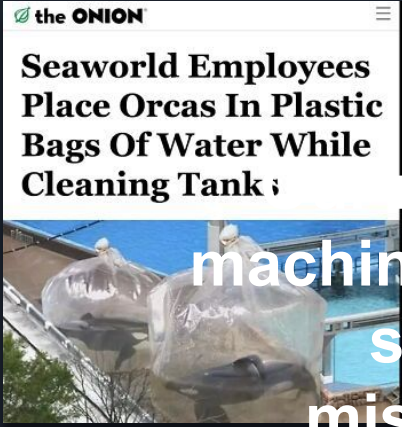
Roadmap of Future Steps

While we are super happy about how far we've come, we recognize that we still have much opportunity for growth ahead of us.

Collaborate with users to create a rating system and feedback loop

Expand to new frontiers of mediums like video or audio

Work with social media platforms to integrate our models directly into their platforms



And this is why I refuse to go to Sea World



We've just overthrown the government of Brazil.

11:52 AM · Nov 10, 2022 · Twitter for iPhone

Linked in

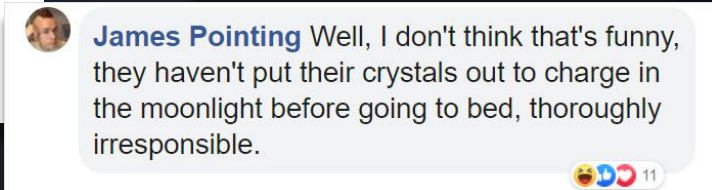


Yesterday I was walking to an interview. There was a starving dog on the road. I stopped to feed him & missed the interview. The next day I got a call asking to come in to do the interview. I was surprised, but I went. Then the interviewer came in. He was the dog



If you are being eaten, please do not scream. You could potentially be spreading covid-19. Keeping visitors safe is our number 1 priority

12:45 PM · Jul 9, 2020



leverage AI to cut out the information from the development of a better digital