



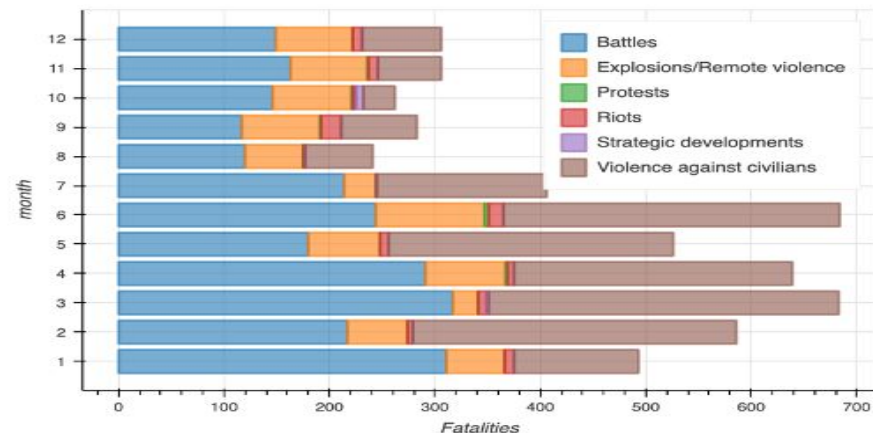
Nigeria News Detective

By Bhargav J Shetgaonkar

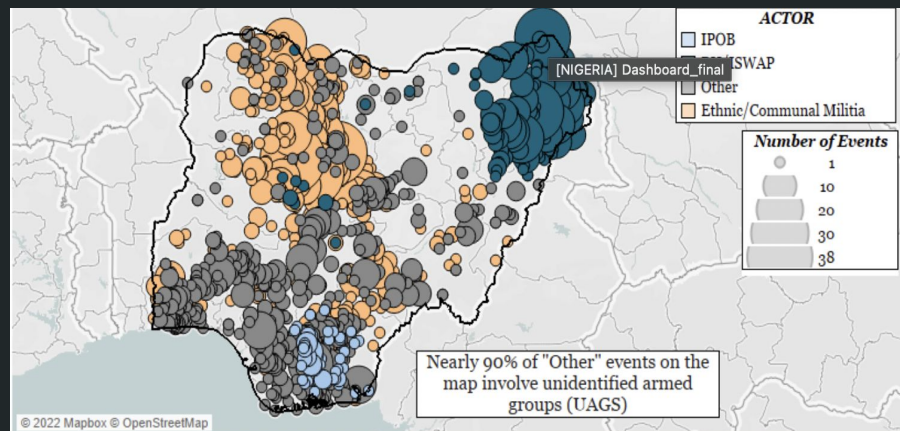
bs337@duke.edu

<https://github.com/BombayBrownBoy09/Nigeria-News-Detective>

Violence in Nigeria 2019 - present



ACLED and the government



These numbers are **alarming!** In order to protect citizens we must monitor & analyze ACLED data

1. Prominent actors include **Boko haram, ISWA,** and similar militia groups
2. Civilian safety is compromised due to **lack of government action**

How can we help?

Identify entities and Inform Citizens accurately

What purpose does ACLED serve?

1. Research tool for media, policy analysts, etc.
2. Information to the public
3. Dashboards for investigation on forecasting events, fatalities, assess risk and plan resources



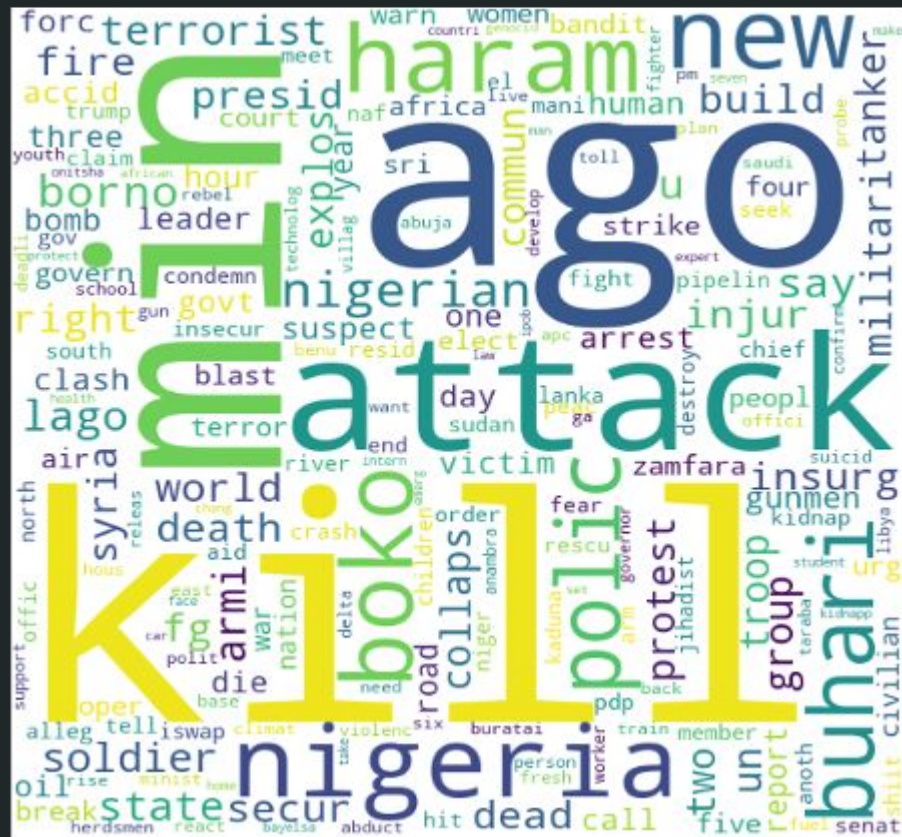
Data Sourcing

Dates used

January 2019 to April 2022

```
Index(['data_id', 'iso', 'event_id_cnty', 'event_id_no_cnty', 'event_date',
      'year', 'time_precision', 'event_type', 'sub_event_type', 'actor1',
      'assoc_actor_1', 'inter1', 'actor2', 'assoc_actor_2', 'inter2',
      'interaction', 'region', 'country', 'admin1', 'admin2', 'admin3',
      'location', 'latitude', 'longitude', 'geo_precision', 'source',
      'source_scale', 'notes', 'fatalities', 'timestamp', 'iso3'],
      dtype='object')
```

*This may be downloaded via the official ACLED
site using your unique access key*



Word cloud representing the notes column

Problem Statement

Using publicly available **ACLED** - The Armed Conflict Location & Event Data to get **news source** and **linked entities** of news articles for government research

of events - approx. **10,000**

Average entities per event **3**

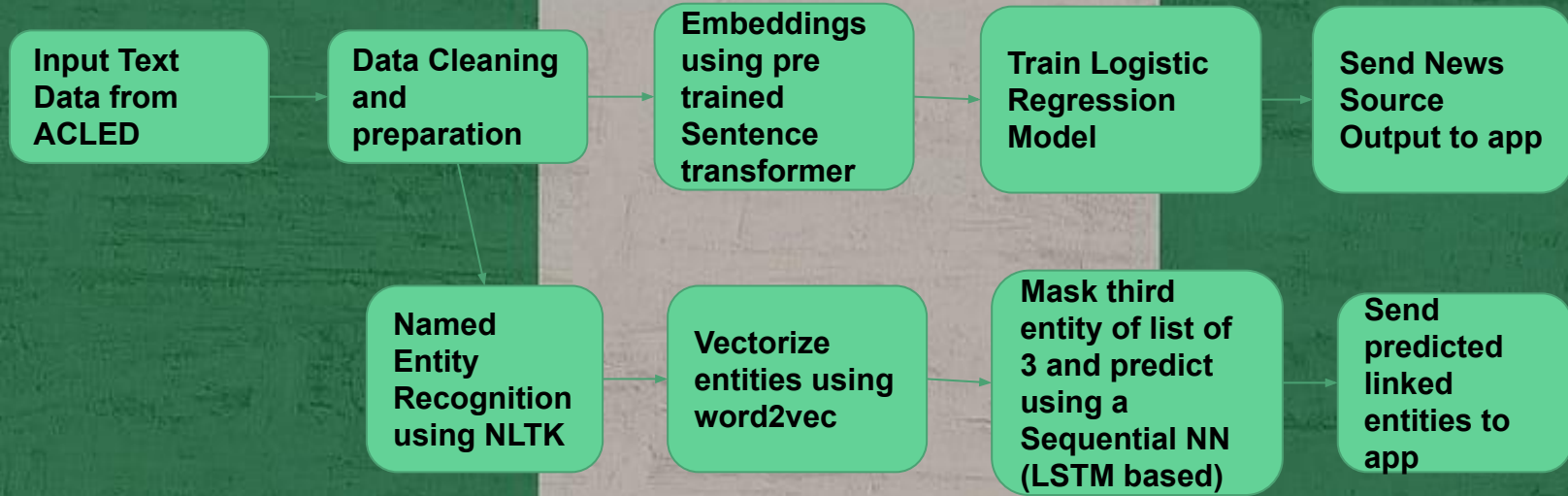
unique news sources **1915**



ACLED

Bringing clarity to crisis

Modeling Pipeline



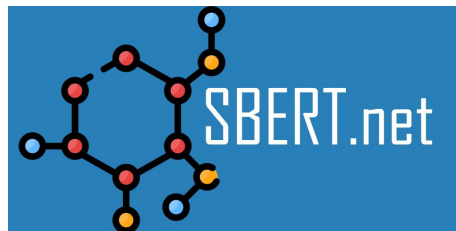
Performance of Models

- Sentence Transformer based **news source classification** - 80/20 training validation split:
Train/Val accuracy - **21% / 16% to classify 1950 types of sources**. Other approaches included **pytorch based** (12% train acc) and **bag of words model** (11.2 % train acc) that performed significantly lower due to the large vocabulary size that results in sparse matrices
- **Entity prediction model** (NN with 2 dense and 2 LSTM layers) had a training / validation split of 75% / 15% and train/validation accuracy of **8%**

```
Model: "sequential"
```

Layer (type)	Output Shape	Param #
lstm (LSTM)	(None, 2, 100)	80400
lstm_1 (LSTM)	(None, 50)	30200
dense (Dense)	(None, 50)	2550
dense_1 (Dense)	(None, 100)	5100

```
=====
Total params: 118,250
Trainable params: 118,250
Non-trainable params: 0
```



Application Demo

Note that these entities are
specific to Nigeria and can be
verified via Google Search



<https://share.streamlit.io/bombaybrownboy09/nigeria-news-detective/main/app.py>

Source and Linked Entities of your Nigeria News

**This is a demo application that employs NLP techniques to predict a news source and linked entities to violent events*

Write or paste news text below

On 1 April 2022, members of a Zamfara militia attacked and killed at least 20 civilians in Daki Takwas village, Gummi LGA (Zamfara) and rustled hundreds of cattle. The police denied that the attack took place.

Submit

Reset

Predicted source: HumAngle

Predicted entities:

1. Ungwan Garama : 0.3324394226074219
2. Kunduru : 0.3021499514579773
3. Zairam Village : 0.27845197916030884
4. Amalgamated Ngbiejeog : 0.272031307220459
5. Anguwar Magaji : 0.26823270320892334

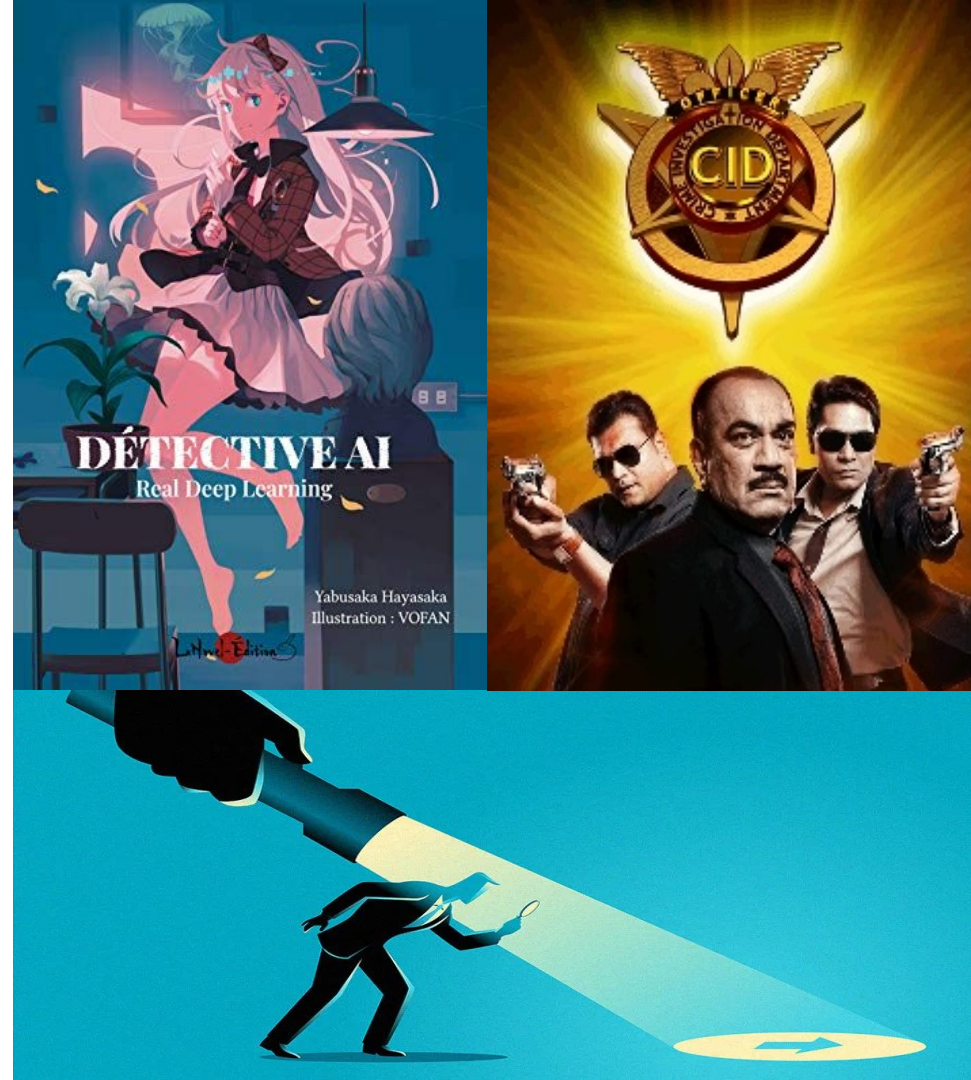
Conclusion

We have used NLP techniques such as text pre processing, named entity recognition, text classification and LSTM based models to build a research tool for media personnel and investigators in Nigeria



What next?

- *Forecasting feature* to predict events
- Improve model performance by *deeper models* and automated updation of data
- *Build UX* for Government Investigators as main admin



Questions?



“Nigeria [in 1990] was all rumour, an unbelievable amount of rumour - largely about crime and almost mythical manifestations of evil.”

- REM KOOLHAAS