# A Comparative Analysis of Semi-Supervised and Self Supervised Classification for Labeling Tweets about Police Brutality

## ABSTRACT

Social media has proven to be influential in social justice advocacy. In 2020, thousands of Nigerians and allies used the EndSARS hashtag to protest police brutality in Nigeria. In this work, we aim to understand the conversation associated with the EndSARS hashtag by comparing the outcome of semi-supervised and self-supervised machine learning classification algorithms for the automatic labeling of tweets. The selfsupervised, zero-shot learning algorithm had the best performance for automatic tweet labeling with average weighted recall of 0.73, compared to cosine similarity with TF-IDF(0.71), cosine similarity with universal sentence encoder (0.58) and Jaccard Similarity(0.222). The major topics of discussion included complaints about police brutality events, Lekki Massacre, activism, media coverage, lack of response from public figures and questions about moving forward.

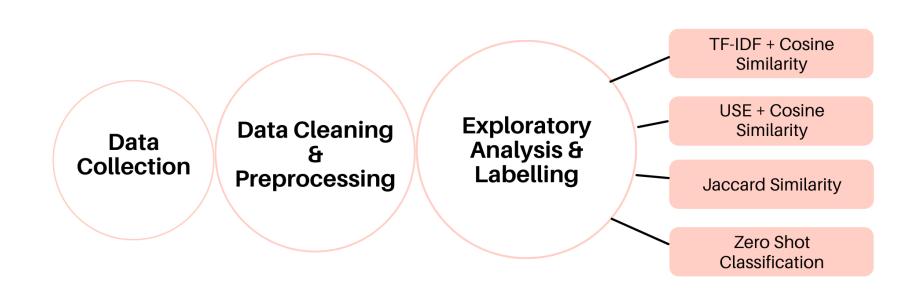
#### Introduction

#### Police brutality and EndSARS

In recent years, there have been a number of protests on the Internet and demonstrations against police around the in-person brutality world. The EndSARS social movement against police brutality which started in 2017 as a Twitter campaign using the hashtag #EndSARS to demand the disbanding of the unit by the Nigerian government, experienced a revitalization in early October 2020. On the night of 20 October 2020, at about 6:50 pm, members of the Nigerian Army were reported to have opened fire on peaceful EndSARS protesters at the Lekki toll gate in Lagos State, Nigeria. Amnesty International stated that at least 12 protesters were killed during the shooting.

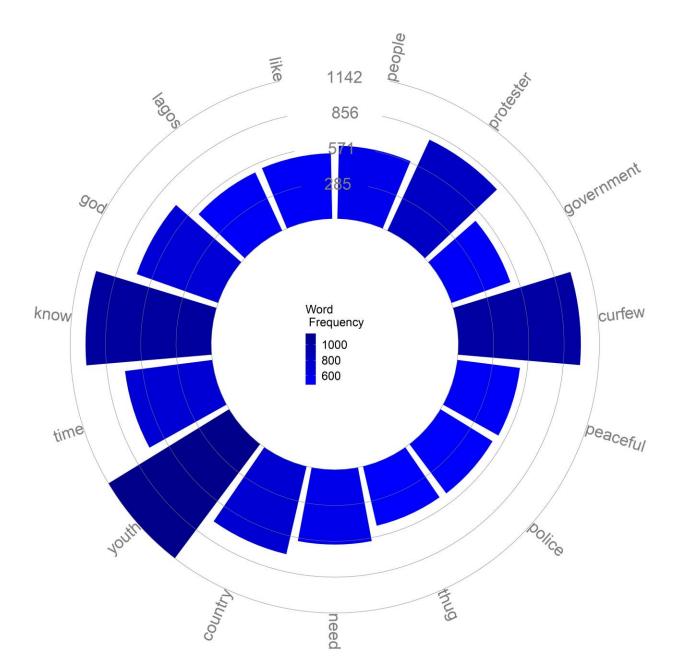
In this work, we analysed tweets related to the EndSARS hashtags, we used unsupervised learning to cluster and perform topic modeling on the tweets to select label topics and present a comparison between semisupervised and self-supervised classification for EndSARS tweets using semantic text similarity and Zero-shot classifier, respectively.

# Methodology



## Exploratory Data Analysis and Label selection

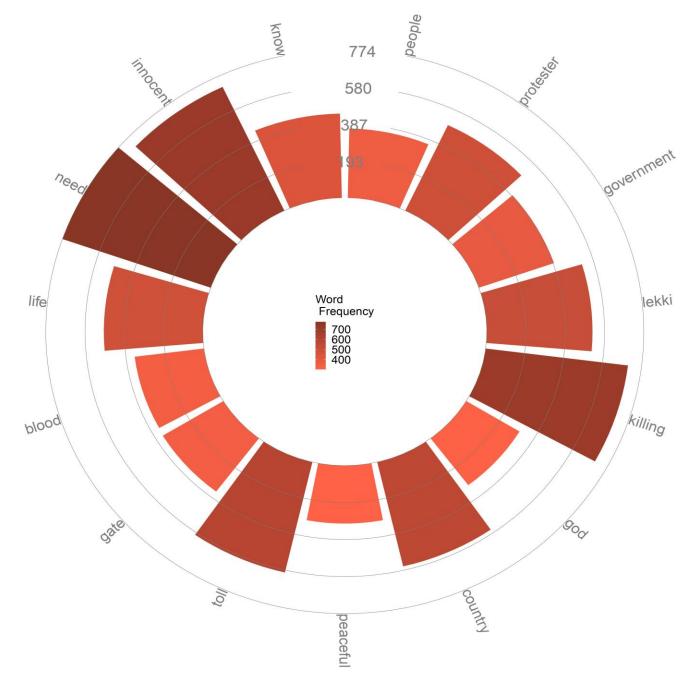
We explored a portion of the dataset that focused on the Lekki massacre event, where Figure 2 shows the most frequent words before 6:50 pm on that day, which reveals tweets about curfew, government, peaceful, protesters etc. Figure 4, on the other hand, shows the word cloud for tweets from 6:50 pm till the end of the day, which reveals tweets about killings, lekki, and blood.



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332,371 tweets with EndSARS hashtags were collected between 1st of October, 2020 to 31st December, 2020 using the Twitter streaming API. We converted all words to

lowercase letters and removed @user, symbols, links, Irrelevant hashtag related tweets, and stopwords



We explored topic modeling for the tweet clustering with LDA model and extracted 6 topic clusters, where each topic has different words that are interconnected. The six topics extracted are as follows: Activism, Police brutality, Media coverage, Lekki massacre, Questions about moving forward/back, No response from public figures.

# **Results and Conclusion**

Table below shows sample tweets and the predicted labels for each method, comparing each of the predicted labels to the human labels,

| s/n | Sample tweet  | Human label  | USE label  | Jaccard label  | TF-IDF label  | Zero-shot label  |
|-----|---|--|--|--|---|--|
| 1   | is there a missing<br>person around you<br>particularly during<br>the movement friends<br>and family should put<br>out their information<br>through the link<br>below | Activism, Questions<br>about moving for-<br>ward/past, Media<br>coverage | Media coverage,<br>Lekki Massacre, Po-<br>lice brutality events                                | Lekki Massacre, Me-<br>dia coverage, Ques-<br>tions about moving<br>forward/past | Police brutality<br>events, Activism,<br>Media coverage             | Activism, Questions<br>about moving for-<br>ward/past, Media<br>coverage |
| 2   | the change begins<br>with you yes you   | Activism   | Activism, No Re-<br>sponse from public<br>figures, Questions<br>about moving for-<br>ward/past | Lekki Massacre, Me-<br>dia coverage, Ques-<br>tions about moving<br>forward/past | Questions about mov-<br>ing forward/past, Ac-<br>tivism, Irrelevant | Activism, Questions<br>about moving for-<br>ward/past, Media<br>coverage |

Zero-shot learning performed better especially with predicting accurate labels for the top 1 prediction. Although for tweets 2 and 3, USE labels were accurate for the top 1 prediction, the send labels are not related to the tweet, e.g., Lekki Massacre label should not be part of the top 3 labels for tweet 3 Jaccard similarity did not perform well on these tweets, since its predictions were almost the same for all tweets. Zero-shot learning proves to be a better approach to providing a more generalized label for multi-label tweets.

| Method                                   | Hamming Loss   | Weighted Average<br>Precision | Weighted Average<br>Recall | Weighted Average<br>F1 score |
|--|----------------|-------------------------------|----------------------------|------------------------------|
| Cosine similarity<br>with TF-IDF         | 0.349          | 0.42                          | 0.71                       | 0.50                         |
| Cosine similarity with USE               | 0.397          | 0.44                          | 0.58                       | 0.44                         |
| Jaccard Similarity<br>Zero-Shot Learning | 0.535<br>0.343 | 0.02<br>0.53                  | 0.22<br><b>0.73</b>        | 0.04<br><b>0.51</b>          |

# References

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