# Content Moderation on Social Media Platforms

Sprucing Up Social Spaces with NLP Magic

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## The problem

### Problem

With the widespread of Social Media influence comes the risk of exposure to harmful content that can negatively impact teen's mental health.

### Solution

Leveraging the power of machine learning to:

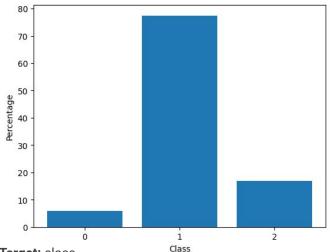
 Automatically detect and flag texts containing cyberbullying, hate speech & other harmful material.

### **Impact**

Make meaningful impact by promoting online safety and well-being for the younger generation.

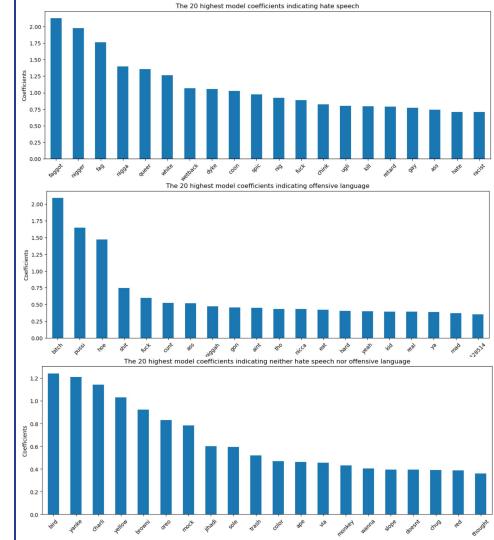
## **EDA**

- 24783 rows & 6 columns
- 0 null values, 0 duplicated rows
- class 0 (6%) hate speech
- class 1 (77%) offensive language
- class 2 (17%) neither



Target: class Feature: tweet





## Preprocessing Procedures

### Train / Test Split

## Split data into train (80%) & test (20%) for modeling.

- The TRAIN set has 19826 data points.
- The TEST set has 4957 data points.

### Re-sampling

## Upsample class 0 and downsample class 1.

- Class 0 before: 1144
- Class 0 after: 3330
- Class 1 before: 15352
- Class 1 after: 3330
- The TRAIN set has 9990 data points.
- The TEST set has 4957 data points.

### Vectorization

## Transform text into numbers.

- Custom tokenizer
- Count Vectorization using Bag-of-Words
- Word Embeddings using Word2Vec

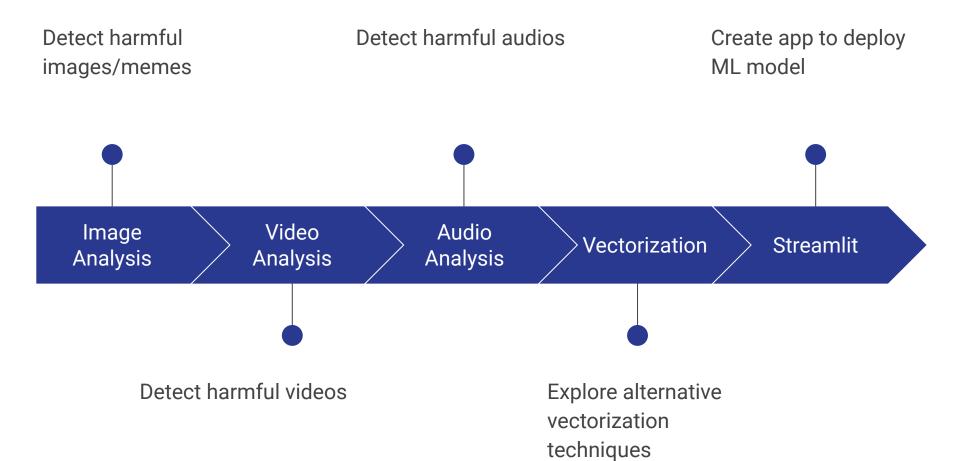
### **Model Selection & Evaluations**

#### Here is our model summary:

Model	Vectorizer	Parameters	Train Accuracy (%)	Test Accuracy (%)	F1-Score (weighted avg)	Recall (weighted avg)
Logistic Regression	CountVectorizer	C=1	83.2	79.3	82	79
Random Forest	Sentence2vecTransformer	max_depth=20, n_estimators=900	99.8	82.1	83	82
XGBoost	Sentence2vecTransformer	max_depth=4, n_estimators=481	99.8	83.6	85	84

	precision	recall	f1-score	support
0	0.35	0.46	0.40	286
1	0.96	0.84	0.90	3838
2	0.64	0.94	0.76	833
accuracy			0.84	4957
macro avg	0.65	0.75	0.69	4957
weighted avg	0.87	0.84	0.85	4957





## Questions