Abstract

This project aims to develop a model that can accurately identify hate speech in text data. By leveraging machine learning algorithms and natural language processing techniques, the objective is to differentiate between hateful and non-hateful content to mitigate the negative impact of hate speech on social platforms. The developed system is trained and tested using a labeled dataset to achieve optimal performance.

Objective

The primary objective of this project is to develop a machine learning model capable of detecting hate speech in textual data. This involves preprocessing text, training an appropriate model, evaluating its performance, and ensuring that it can be effectively used for hate speech moderation.

Introduction

Hate speech is a form of expression that can promote discrimination and violence, especially on social media platforms. As the online community grows, the need for automated tools to detect and mitigate hate speech becomes increasingly vital. This project addresses this need by developing a hate speech detection model using machine learning and natural language processing.

Methodology

Data Collection: A labeled dataset containing examples of hate speech and non-hate speech was used for training and testing. Data Preprocessing: Text data was cleaned by removing punctuation, converting to lowercase, and removing stop words. Feature Extraction: Techniques like TF-IDF or word embeddings were employed to transform the text into a numerical form suitable for model training. Model Training: Various machine learning models (e.g., logistic regression, SVM, or deep learning) were trained using the preprocessed data. Evaluation: The models were evaluated using metrics like accuracy, precision, recall, and F1-score.

import warnings
warnings.filterwarnings("ignore")

In [1]:

import numpy as np
import pandas as pd

dataset=pd.read_csv("C:\\Users\\ASUS\\Downloads\\twitter_data.csv")
dataset

	Unnamed: 0	count	hate_speech	offensive_language	neither	class	tweet
0	0	3	0	0	3		!!! RT @mayasolovely: As a woman you shouldn't

Out[3]:

	Unnamed:	count	hate_speech	offensive_language	neither	class	tweet
1	1	3	0	3	0	1	!!!!! RT @mleew17: boy dats coldtyga dwn ba
2	2	3	0	3	0	1	!!!!!!! RT @UrKindOfBrand Dawg!!!! RT @80sbaby
3	3	3	0	2	1	1	!!!!!!!!! RT @C_G_Anderson: @viva_based she lo
4	4	6	0	6	0	1	!!!!!!!!!!!! RT @ShenikaRoberts: The shit you
•••	•••						
24778	25291	3	0	2	1	1	you's a muthaf***in lie "@LifeAsKing: @2
24779	25292	3	0	1	2	2	you've gone and broke the wrong heart baby, an
24780	25294	3	0	3	0	1	young buck wanna eat!! dat nigguh like I ain
24781	25295	6	0	6	0	1	youu got wild bitches tellin you lies
24782	25296	3	0	0	3	2	~~Ruffled Ntac Eileen Dahlia - Beautiful col

24783 rows × 7 columns

dataset.isnull().sum()

Out[4]:

In [4]:

Unnamed: 0 0 count 0 hate_speech 0 offensive_language 0 neither 0 class 0 tweet 0

dtype: int64

In [5]:

dataset.info()

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 24783 entries, 0 to 24782

Data columns (total 7 columns):

Column Non-Null Count Dtype

O Unnamed: 0 24783 non-null int64

count 24783 non-null int64

2 hate_speech 24783 non-null int64
3 offensive_language 24783 non-null int64
4 neither 24783 non-null int64
5 class 24783 non-null int64
6 tweet 24783 non-null object

dtypes: int64(6), object(1)

memory usage: 1.3+ MB

dataset.describe()

In [6]:

Out[6]:

						σατ[σ].
	Unnamed: 0	count	hate_speech	offensive_language	neither	class
count	24783.000000	24783.000000	24783.000000	24783.000000	24783.000000	24783.000000
mean	12681.192027	3.243473	0.280515	2.413711	0.549247	1.110277
std	7299.553863	0.883060	0.631851	1.399459	1.113299	0.462089
min	0.000000	3.000000	0.000000	0.000000	0.000000	0.000000
25%	6372.500000	3.000000	0.000000	2.000000	0.000000	1.000000
50%	12703.000000	3.000000	0.000000	3.000000	0.000000	1.000000
75 %	18995.500000	3.000000	0.000000	3.000000	0.000000	1.000000
max	25296.000000	9.000000	7.000000	9.000000	9.000000	2.000000

In [7]:

dataset["labels"]=dataset["class"].map({0:"Hate Speech",1:"Offensive",2: "No
hate or offensive"})

In [8]:

dataset

Out[8]:

	Unnamed: 0	count	hate_speech	offensive_language	neither	class	tweet	labels
0	0	3	0	0	3	2	!!! RT @mayasolovely: As a woman you shouldn't	No hate or offensive
1	1	3	0	3	0	1	!!!!! RT @mleew17: boy dats coldtyga dwn ba	Offensive
2	2	3	0	3	0	1	!!!!!!! RT @UrKindOfBrand Dawg!!!! RT @80sbaby	Offensive
3	3	3	0	2	1	1	!!!!!!!!! RT @C_G_Anderson: @viva_based she lo	Offensive
4	4	6	0	6	0	1	!!!!!!!!!!! RT @ShenikaRoberts: The shit you	Offensive
•••								
24778	25291	3	0	2	1	1	you's a muthaf***in lie හ#8220;@LifeAsKing: @2	Offensive

	Unnamed:	count	hate_speech	offensive_language	neither	class	tweet	labels
24779	25292	3	0	1	2	2	you've gone and broke the wrong heart baby, an	
24780		3	0	3	0	1	young buck wanna eat!! dat nigguh like I ain	Offensive
24781	25295	6	0	6	0		youu got wild bitches tellin you lies	Offensive
24782		3	0	0	3		T.1 5 1 1.	No hate or offensive

24783 rows × 8 columns

data=dataset[["tweet", "labels"]]

In [9]: In [10]:

data

Out[10]:

	tweet	labels
0	!!! RT @mayasolovely: As a woman you shouldn't	No hate or offensive
1	!!!!! RT @mleew17: boy dats coldtyga dwn ba	Offensive
2	!!!!!!! RT @UrKindOfBrand Dawg!!!! RT @80sbaby	Offensive
3	!!!!!!!!! RT @C_G_Anderson: @viva_based she lo	Offensive
4	!!!!!!!!!!! RT @ShenikaRoberts: The shit you	Offensive
•••		•••
24778	you's a muthaf***in lie "@LifeAsKing: @2	Offensive
24779	you've gone and broke the wrong heart baby, an	No hate or offensive
24780	young buck wanna eat!! dat nigguh like I ain	Offensive
24781	youu got wild bitches tellin you lies	Offensive
24782	~~Ruffled Ntac Eileen Dahlia - Beautiful col	No hate or offensive

24783 rows × 2 columns

True

In [11]:

```
import re
import nltk
import string
nltk.download('stopwords')
[nltk_data] Downloading package stopwords to
                C:\Users\ASUS\AppData\Roaming\nltk_data...
[nltk_data]
[nltk_data]
              Package stopwords is already up-to-date!
```

Out[11]:

from nltk.corpus import stopwords

In [12]:

In [13]:

stopwords=set(stopwords.words("english"))

Data Cleaning:

```
In [14]:
def clean(text):
    text = str(text).lower()
    text = re.sub('\[.*?\]', '', text)
    text = re.sub('https?://\S+|www\.\S+', '', text)
    text = re.sub('<.*?>+', '', text)
    text = re.sub(r"\@w+|\#",'',text)
    text = re.sub(r"[^\w\s]",'',text)
    text = re.sub('[%s]' % re.escape(string.punctuation), '', text)
    text = re.sub('\n', '', text)
    text = re.sub('\w^*\d\w^*', '', text)
    text = [word for word in text.split(' ') if word not in stopwords]
#removing stopwords
    text = " ".join(text)
    text = [Stemmer.stem(word) for word in text.split(' ')]
    text = " ".join(text)
    return text
                                                                          In [15]:
data['tweet'] = data['tweet'].apply(clean)
data
```

Out[15]:

	tweet	labels
0	rt mayasolov woman shouldnt complain clean ho	No hate or offensive
1	rt boy dat coldtyga dwn bad cuffin dat hoe	Offensive
2	rt urkindofbrand dawg rt ever fuck bitch sta	Offensive
3	rt cganderson vivabas look like tranni	Offensive
4	rt shenikarobert shit hear might true might f	Offensive
•••		•••
24778	yous muthafin lie coreyemanuel right tl tras	Offensive
24779	youv gone broke wrong heart babi drove redneck	No hate or offensive
24780	young buck wanna eat dat nigguh like aint fuck	Offensive
24781	youu got wild bitch tellin lie	Offensive
24782	ruffl ntac eileen dahlia beauti color combin	No hate or offensive

 $24783 \text{ rows} \times 2 \text{ columns}$

```
In [16]:
X=np.array(data['tweet'])
y=np.array(data['labels'])

In [17]:
X
Out[17]:
```

```
array([' rt mayasolov woman shouldnt complain clean hous amp man alway take
trash',
       'rt boy dat coldtyga dwn bad cuffin dat hoe place',
       'rt urkindofbrand dawg rt ever fuck bitch start cri confus shit',
       ..., 'young buck wanna eat dat nigguh like aint fuckin dis',
       'youu got wild bitch tellin lie',
       'ruffl ntac eileen dahlia beauti color combin pink orang yellow amp
white coll '],
      dtype=object)
                                                                          In [18]:
from sklearn.feature extraction.text import CountVectorizer
from sklearn.model_selection import train_test_split
                                                                          In [19]:
cv =CountVectorizer()
X=cv.fit_transform(X)
                                                                          In [20]:
Χ
                                                                          Out[20]:
<24783x25697 sparse matrix of type '<class 'numpy.int64'>'
       with 197861 stored elements in Compressed Sparse Row format>
                                                                          In [21]:
X_train ,X_test,y_train,y_test=train_test_split(X,y,test_size=0.33,random_sta
te=42)
                                                                          In [22]:
X_train
                                                                          Out[22]:
<16604x25697 sparse matrix of type '<class 'numpy.int64'>'
       with 132620 stored elements in Compressed Sparse Row format>
                                                                          In [23]:
from sklearn.tree import DecisionTreeClassifier
                                                                          In [24]:
dt=DecisionTreeClassifier()
dt.fit(X_train,y_train)
                                                                          Out[24]:
DecisionTreeClassifier()
                                                                          In [25]:
y_pred=dt.predict(X_test)
                                                                          In [26]:
from sklearn.metrics import confusion_matrix
cm=confusion_matrix(y_test,y_pred)
cm
                                                                          Out[26]:
array([[ 154, 38, 273],
       [ 33, 1161, 185],
       [ 227, 256, 5852]], dtype=int64)
                                                                          In [27]:
import seaborn as sns
```

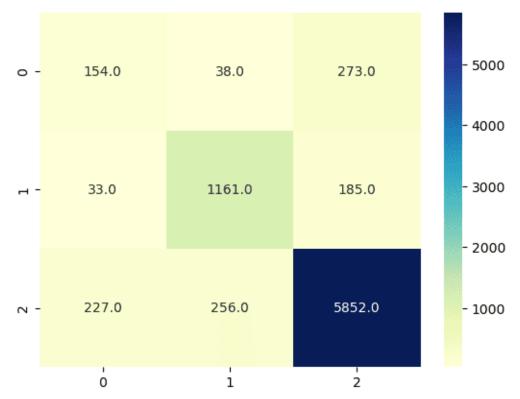
import matplotlib.pyplot as plt
%matplotlib inline

sns.heatmap(cm, annot = True, fmt=".1f", cmap="YlGnBu")

In [28]:

Out[28]:

<AxesSubplot:>



from sklearn.metrics import accuracy_score
accuracy_score(y_test,y_pred)

Out[29]:

In [29]:

0.8762684924807433

Model Sample:

sample="kill all the people"
sample=clean(sample)

In [31]:

In [30]:

 ${\tt sample}$

Out[31]:

'kill peopl'

data1=cv.transform([sample]).toarray()

In [32]:

data1

In [33]:

```
Out[33]:
array([[0, 0, 0, ..., 0, 0, 0]], dtype=int64)

In [34]:
dt.predict(data1)

Out[34]:
```

Conclusion

The project successfully developed a model that can classify text as hate speech or non-hate speech with reasonable accuracy. The performance metrics indicate that the model can be a helpful tool for moderating online content, but further improvement can be achieved by increasing the dataset size and exploring more advanced models.

In []: