ChatGPT Sentiment Analysis Using ML and NLP

```
In [1]: #importing the Dependinces
        from nltk.util import pr
        import matplotlib.pyplot as plt
        import pandas as pd
        import numpy as np
        from sklearn.feature_extraction.text import CountVectorizer
        from sklearn.model_selection import train_test_split
        from sklearn.tree import DecisionTreeClassifier
        import re
        import nltk
        stemmer = nltk.SnowballStemmer("english")
        from nltk.corpus import stopwords
        import string
        stopword=set(stopwords.words('english'))
In [2]: | nltk.download('stopwords')
        [nltk_data] Downloading package stopwords to /Users/vikky/nltk_dat
        [nltk_data]
                      Package stopwords is already up-to-date!
Out[2]: True
```

Data collection

In [3]: #data loading in pandas

```
In [4]: #check first five rows
data.head()
```

Out[4]:

labels	tweets	Unnamed: 0	
neutral	ChatGPT: Optimizing Language Models for Dialog	0	0
good	Try talking with ChatGPT, our new Al system wh	1	1
neutral	ChatGPT: Optimizing Language Models for Dialog	2	2
good	THRILLED to share that ChatGPT, our new model	3	3
bad	As of 2 minutes ago, @OpenAl released their ne	4	4

```
In [5]: #check last five rows
        data.tail()
```

Out [5]:

labels	tweets	Unnamed: 0	
bad	Other Software Projects Are Now Trying to Repl	219289	219289
good	I asked #ChatGPT to write a #NYE Joke for SEOs	219290	219290
bad	chatgpt is being disassembled until it can onl	219291	219291
bad	2023 predictions by #chatGPT. Nothing really s	219292	219292
neutral	From ChatGPT, neat stuff https://t.co/qjjUF2Z2m0	219293	219293

```
In [6]: #check shape
        data.shape
```

Out[6]: (219294, 3)

In [7]: #check more infomation data.info()

<class 'pandas.core.frame.DataFrame'> RangeIndex: 219294 entries, 0 to 219293 Data columns (total 3 columns):

```
#
    Column
                 Non-Null Count
                                  Dtype
    Unnamed: 0 219294 non-null
                                  int64
 1
    tweets
                 219294 non-null
                                  object
 2
                 219294 non-null
     labels
                                  object
dtypes: int64(1), object(2)
memory usage: 5.0+ MB
```

In [8]: #check missing value data.isnull().sum()

Out[8]: Unnamed: 0 0 tweets 0 labels 0 dtype: int64

```
In [9]: #check duplicated value in data set
        data.duplicated().sum()
```

Out[9]: 0

```
In [10]: #check unused columns
         del data['Unnamed: 0']
```

```
In [11]: #clean data set
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('[%s]' % re.escape(string.punctuation), '', text)
    text = re.sub('\n', '', text)
    text = re.sub('\w*\d\w*', '', text)
    text = re.sub('\w*\d\w*', '', text)
    text = [word for word in text.split(' ') if word not in stopword
    text=" ".join(text)
    text = [stemmer.stem(word) for word in text.split(' ')]
    text=" ".join(text)
    return text
data["tweets"] = data["tweets"].apply(clean)
```

```
In [12]: #after check clean dataset
    data.head()
```

Out[12]:

	tweets	labels	
0	chatgpt optim languag model dialogu openai	neutral	
1	tri talk chatgot new ai system optim dialogu f	good	

- 2 chatgpt optim languag model dialogu ai machin... neutral
- 3 thrill share chatgpt new model optim dialog pu... good
- 4 minut ago openai releas new chatgpt nnand use... bad

```
In [13]: #check value of labels
data['labels'].value_counts()
```

Out[13]: bad 107796 good 56011 neutral 55487

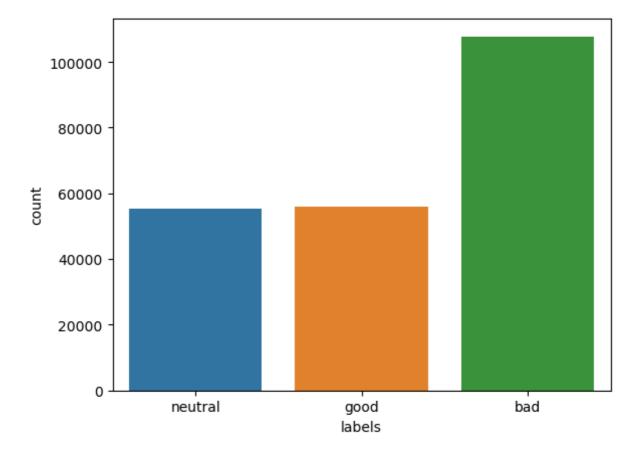
Name: labels, dtype: int64

In [14]: import seaborn as sns
sns.countplot(data['labels'])

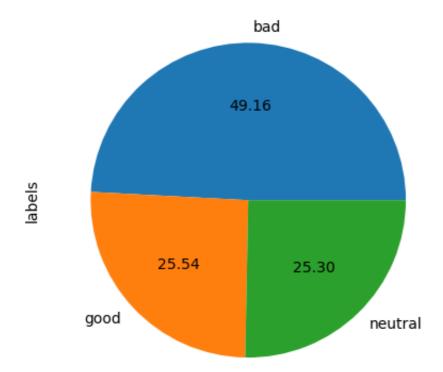
/Users/vikky/opt/anaconda3/lib/python3.9/site-packages/seaborn/_de corators.py:36: FutureWarning: Pass the following variable as a ke yword arg: x. From version 0.12, the only valid positional argumen t will be `data`, and passing other arguments without an explicit keyword will result in an error or misinterpretation.

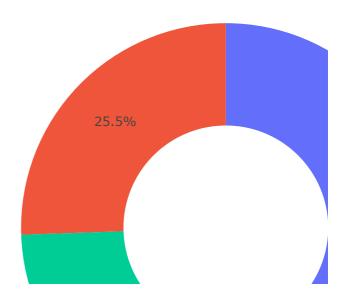
warnings.warn(

Out[14]: <AxesSubplot:xlabel='labels', ylabel='count'>



```
In [15]: data['labels'].value_counts().plot(kind='pie',autopct='%.2f')
Out[15]: <AxesSubplot:ylabel='labels'>
```





In [17]: from nltk.sentiment.vader import SentimentIntensityAnalyzer
from wordcloud import WordCloud, STOPWORDS, ImageColorGenerator

In [18]: !pip install wordcloud

Requirement already satisfied: wordcloud in /Users/vikky/opt/anaconda3/lib/python3.9/site-packages (1.8.2.2)

Requirement already satisfied: pillow in /Users/vikky/opt/anaconda 3/lib/python3.9/site-packages (from wordcloud) (9.2.0)

Requirement already satisfied: matplotlib in /Users/vikky/opt/anac onda3/lib/python3.9/site-packages (from wordcloud) (3.5.2)

Requirement already satisfied: numpy>=1.6.1 in /Users/vikky/opt/an aconda3/lib/python3.9/site-packages (from wordcloud) (1.21.5)

Requirement already satisfied: python-dateutil>=2.7 in /Users/vikk y/opt/anaconda3/lib/python3.9/site-packages (from matplotlib->word cloud) (2.8.2)

Requirement already satisfied: kiwisolver>=1.0.1 in /Users/vikky/o pt/anaconda3/lib/python3.9/site-packages (from matplotlib->wordclo ud) (1.4.2)

Requirement already satisfied: pyparsing>=2.2.1 in /Users/vikky/op t/anaconda3/lib/python3.9/site-packages (from matplotlib->wordclou d) (3.0.9)

Requirement already satisfied: packaging>=20.0 in /Users/vikky/opt/anaconda3/lib/python3.9/site-packages (from matplotlib->wordcloud) (21.3)

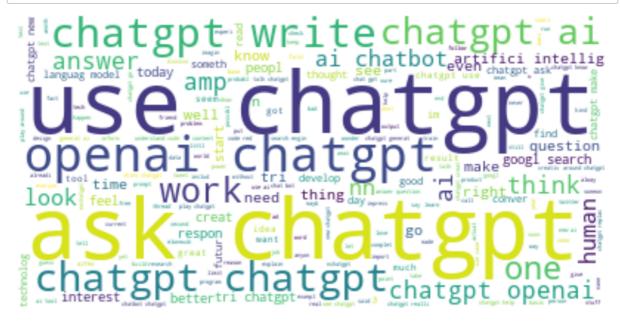
Requirement already satisfied: fonttools>=4.22.0 in /Users/vikky/o pt/anaconda3/lib/python3.9/site-packages (from matplotlib->wordclo ud) (4.25.0)

Requirement already satisfied: cycler>=0.10 in /Users/vikky/opt/an aconda3/lib/python3.9/site-packages (from matplotlib->wordcloud) (0.11.0)

Requirement already satisfied: six>=1.5 in /Users/vikky/opt/anacon da3/lib/python3.9/site-packages (from python-dateutil>=2.7->matplo tlib->wordcloud) (1.16.0)

[notice] A new release of pip available: 22.3.1 -> 23.0.1
[notice] To update, run: pip install --upgrade pip





```
In [21]: #spliting the dataset
X = data['tweets']
Y = data['labels']
```

In [22]: #loading CountVectorizer
cv = CountVectorizer()
X = cv.fit_transform(X)

In [23]: print(X)

```
(0, 18069)
               1
(0, 87811)
               1
(0.63492)
               1
(0, 73132)
               1
(0, 31060)
               1
(0, 87253)
               1
(1, 18069)
               1
(1, 87811)
               1
(1, 31060)
               1
(1, 122661)
               1
(1, 115953)
               1
(1, 77580)
               1
(1, 2194)
               1
(1, 115594)
```

(1, 40734)

(1, 51059)

1

1

```
(1, 125978)
                           1
            (1, 55151)
                           1
            (2, 18069)
                           1
            (2, 87811)
                           1
            (2, 63492)
                           1
            (2, 73132)
                           1
            (2, 31060)
                           1
            (2, 2194)
                           1
            (2, 67874)
                           1
            (219290, 79731)
                                    1
            (219290, 82867)
                                    1
            (219290, 85574)
                                    1
                                    1
            (219290, 49448)
            (219290, 50721)
                                    1
                                    1
            (219290, 77887)
            (219291, 18069)
                                    1
            (219291, 31835)
                                    1
            (219291, 32196)
                                    1
            (219292, 18069)
                                    1
                                    1
            (219292, 98853)
            (219292, 106117)
                                    1
            (219292, 90006)
                                    1
            (219292, 133676)
                                    1
            (219292, 6922)
                                    1
                                    1
            (219292, 28648)
            (219292, 122588)
                                    1
                                    1
            (219292, 121700)
            (219292, 94110)
                                    1
            (219292, 83814)
                                    1
            (219292, 111482)
                                    1
            (219292, 84729)
                                    1
                                    1
            (219293, 18069)
            (219293, 113876)
                                    1
            (219293, 76907)
                                    1
In [24]: from sklearn.preprocessing import LabelEncoder
          le = LabelEncoder()
          Y = le.fit_transform(Y)
In [25]: | print(Y)
          [2 1 2 ... 0 0 2]
In [26]: #spliting the dataset
         X_train, X_test, Y_train, Y_test = train_test_split(X, Y, test_size=
```

```
In [27]: #Check shape of X_train and X_test or Y_train
         print(X_train.shape, X_test.shape, Y_train.shape)
         (146926, 135976) (72368, 135976) (146926,)
In [28]: from sklearn.linear_model import LogisticRegression
         lg = LogisticRegression()
         lg.fit(X_train, Y_train)
         /Users/vikky/opt/anaconda3/lib/python3.9/site-packages/sklearn/lin
         ear model/ logistic.py:458: ConvergenceWarning:
         lbfgs failed to converge (status=1):
         STOP: TOTAL NO. of ITERATIONS REACHED LIMIT.
         Increase the number of iterations (max_iter) or scale the data as
         shown in:
             https://scikit-learn.org/stable/modules/preprocessing.html
         (https://scikit-learn.org/stable/modules/preprocessing.html)
         Please also refer to the documentation for alternative solver opti
         ons:
             https://scikit-learn.org/stable/modules/linear_model.html#logi
         stic-regression (https://scikit-learn.org/stable/modules/linear mo
         del.html#logistic-regression)
Out [28]:
          ▼ LogisticRegression
         LogisticRegression()
In [29]: | from sklearn.metrics import accuracy_score
         #traing dataset accuracy score
         train_test = lg.predict(X_train)
         accuracy_score(train_test,Y_train)
Out [29]: 0.9091651579706791
In [30]: #test dataset accuracy score
         test data = lq.predict(X test)
         accuracy_score(test_data, Y_test)
```

Out[30]: 0.8460092858722087

In [31]: from sklearn.metrics import jaccard_score, accuracy_score, f1_score,
 preds = lg.predict(X_test)
 print(classification_report(Y_test, preds))

	precision	recall	f1-score	support
0 1 2	0.89 0.86 0.74	0.93 0.83 0.70	0.91 0.84 0.72	35518 18508 18342
accuracy macro avg weighted avg	0.83 0.84	0.82 0.85	0.85 0.82 0.84	72368 72368 72368

- In [33]: #using from sklearn.tree import DecisionTreeClassifier
 from sklearn.tree import DecisionTreeClassifier
 DT = DecisionTreeClassifier()
- In [34]: #model fit
 DT.fit(X_train,Y_train)
- Out[34]: v DecisionTreeClassifier DecisionTreeClassifier()
- In [35]: #traing dataset accuracy score
 train_test = DT.predict(X_train)
 accuracy_score(train_test,Y_train)
- Out[35]: 0.9992921606795258
- In [36]: #test dataset accuracy score
 test_data = DT.predict(X_test)
 accuracy_score(test_data, Y_test)
- Out[36]: 0.787862038470042

```
In [37]: #classification report
         preds = DT.predict(X test)
         print(classification_report(Y_test, preds))
                                     recall f1-score
                        precision
                                                         support
                                       0.87
                     0
                             0.89
                                                 0.88
                                                           35518
                             0.75
                                                 0.74
                     1
                                       0.73
                                                           18508
                     2
                             0.64
                                       0.68
                                                 0.66
                                                           18342
                                                 0.79
                                                           72368
             accuracy
                             0.76
                                       0.76
                                                 0.76
                                                           72368
            macro avq
                                                 0.79
                                                           72368
         weighted avg
                             0.79
                                       0.79
In [43]: from sklearn.naive_bayes import MultinomialNB
         nb = MultinomialNB()
         nb.fit(X_train, Y_train)
Out[43]:
          ▼ MultinomialNB
         MultinomialNB()
In [44]: preds = nb.predict(X_test)
         print(classification_report(Y_test, preds))
                        precision
                                     recall f1-score
                                                         support
                             0.80
                                       0.89
                                                 0.84
                     0
                                                           35518
                     1
                             0.63
                                       0.81
                                                 0.71
                                                           18508
                                       0.30
                             0.58
                                                 0.39
                     2
                                                           18342
                                                 0.72
                                                           72368
             accuracy
                             0.67
                                       0.66
                                                 0.65
                                                           72368
            macro avq
                             0.70
                                       0.72
                                                 0.70
                                                           72368
         weighted avg
In [45]: #Hyperparameter tuning
         from sklearn.model_selection import GridSearchCV, RepeatedStratified
In [46]: # Hyperparameter tuning for Multinomial Naive Bayes model
         param_grid = {"alpha": [0.1,0,1.0, 10, 100]}
         grid_search = GridSearchCV(MultinomialNB(), param_grid, verbose=2)
         grid_search.fit(X_train, Y_train)
```

Fitting 5 folds for each of 5 candidates, totalling 25 fits

[CV] ENDalpha=0.1; tota

l time= 0.1s	
[CV] ENDalpha=0.1;	tota
l time= 0.1s	
[CV] ENDalpha=0.1;	tota
l time= 0.1s	
[CV] ENDalpha=0.1;	tota
l time= 0.1s	
[CV] ENDalpha=0.1;	tota
l time= 0.1s	
[CV] ENDalpha=0;	tota
l time= 0.1s	
[CV] ENDalpha=0;	tota
l time= 0.1s	

/Users/vikky/opt/anaconda3/lib/python3.9/site-packages/sklearn/naive_bayes.py:627: FutureWarning:

The default value for `force_alpha` will change to `True` in 1.4. To suppress this warning, manually set the value of `force_alpha`.

/Users/vikky/opt/anaconda3/lib/python3.9/site-packages/sklearn/naive_bayes.py:633: UserWarning:

alpha too small will result in numeric errors, setting alpha = 1.0 e-10. Use `force_alpha=True` to keep alpha unchanged.

/Users/vikky/opt/anaconda3/lib/python3.9/site-packages/sklearn/naive_bayes.py:627: FutureWarning:

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alpha too small will result in numeric errors, setting alpha = 1.0 e-10. Use `force_alpha=True` to keep alpha unchanged.

[CV] END	alpha=0;	tota
l time=	0.1s	
[CV] END	alpha=0:	tota

l time= 0.1s

/Users/vikky/opt/anaconda3/lib/python3.9/site-packages/sklearn/naive_bayes.py:627: FutureWarning:

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[CV] END	alpha=0;	tota
[CV] END	alpha=1.0;	tota
[CV] END	alpha=1.0;	tota
[CV] END	alpha=1.0;	tota
l time= 0.1s [CV] END	alpha=1.0;	tota
l time= 0.3s [CV] END	alpha=1.0;	tota
l time= 0.1s [CV] END	alpha=10;	tota
l time= 0.1s [CV] END	alpha=10;	tota
l time= 0.2s [CV] END	alpha=10;	tota
l time= 0.2s [CV] END	alpha=10;	tota
l time= 0.2s [CV] END	•	
l time= 0.2s [CV] END	•	
l time= 0.3s [CV] END	•	
l time= 0.2s [CV] END	•	
[CV] END	a tpiia-100;	tuta