# prodigy-ds-04

#### December 18, 2023

Task-04: Analyze and visualize sentiment patterns in social media data to understand public opinion and attitudes towards specific topics or brands.

Problem Statement: A Twitter sentiment analysis uses NLP and ML models to classify tweets into negative, positive or neutral emotions.

```
[]: import numpy as np
    import pandas as pd
    import seaborn as sns
    import matplotlib.pyplot as plt
    from sklearn.preprocessing import LabelEncoder
    from sklearn.feature_extraction.text import TfidfVectorizer
    from sklearn.model_selection import train_test_split
    from sklearn.linear_model import LogisticRegression
    from sklearn.metrics import
      Glassification_report,accuracy_score,confusion_matrix
    from mlxtend.plotting import plot confusion matrix
    from sklearn.naive_bayes import MultinomialNB
    import nltk
    from nltk.sentiment.vader import SentimentIntensityAnalyzer
[]: from google.colab import files
    data = files.upload()
    <IPython.core.display.HTML object>
    Saving twitter_training.csv to twitter_training.csv
[]: df=pd.read csv('twitter training.csv')
[]: df
[]:
            2401 Borderlands Positive \
            2401 Borderlands Positive
    0
    1
            2401 Borderlands Positive
    2
            2401 Borderlands Positive
    3
            2401 Borderlands Positive
    4
            2401 Borderlands Positive
```

```
74676
           9200
                       Nvidia Positive
     74677
            9200
                       Nvidia Positive
     74678
            9200
                       Nvidia Positive
     74679
            9200
                       Nvidia Positive
     74680
           9200
                       Nvidia Positive
           im getting on borderlands and i will murder you all ,
     0
            I am coming to the borders and I will kill you...
     1
            im getting on borderlands and i will kill you ...
     2
            im coming on borderlands and i will murder you...
     3
            im getting on borderlands 2 and i will murder ...
     4
            im getting into borderlands and i can murder y...
           Just realized that the Windows partition of my...
    74676
     74677
            Just realized that my Mac window partition is ...
     74678
            Just realized the windows partition of my Mac ...
     74679
            Just realized between the windows partition of...
     74680
            Just like the windows partition of my Mac is 1...
     [74681 rows x 4 columns]
[]: df.columns=['id', 'country', 'label', 'text']
[]: df2=df.copy()
[]: df.shape
[]: (74681, 4)
[]: df.head()
[]:
                              label \
          id
                  country
       2401 Borderlands Positive
     0
     1 2401 Borderlands Positive
     2 2401 Borderlands Positive
     3 2401 Borderlands Positive
     4 2401 Borderlands Positive
                                                      text
     O I am coming to the borders and I will kill you...
     1 im getting on borderlands and i will kill you ...
     2 im coming on borderlands and i will murder you...
     3 im getting on borderlands 2 and i will murder ...
     4 im getting into borderlands and i can murder y...
[]: df.tail()
```

```
[]:
              id country
                             label \
    74676
           9200
                 Nvidia Positive
    74677
            9200
                 Nvidia Positive
     74678
           9200
                 Nvidia Positive
     74679
            9200
                 Nvidia Positive
     74680
           9200
                 Nvidia Positive
                                                          text
     74676
            Just realized that the Windows partition of my...
     74677
            Just realized that my Mac window partition is ...
     74678
            Just realized the windows partition of my Mac ...
     74679
            Just realized between the windows partition of...
     74680
            Just like the windows partition of my Mac is 1...
[]: df.iloc[100:200]
[]:
            id
                    country
                                  label \
     100
         2417
                Borderlands
                               Negative
     101
         2418
               Borderlands
                             Irrelevant
     102 2418
               Borderlands
                             Irrelevant
     103 2418
                             Irrelevant
               Borderlands
     104 2418
               Borderlands
                             Irrelevant
     . .
     195
         2433
               Borderlands
                                Neutral
     196 2433 Borderlands
                                Neutral
     197 2434 Borderlands
                               Negative
     198 2434 Borderlands
                               Negative
     199 2434 Borderlands
                               Negative
                                                        text
     100 Grounded almost was pretty cool even despite t...
     101
          Appreciate the (sonic) concepts / praxis Valen...
     102 Appreciate the (sound) concepts / practices th...
     103 Evaluate the (sound) concepts / concepts of Va...
     104
         Appreciate the (sonic) concepts / praxis Valen...
     . .
     195
          i then enter in that gunner seat and i fear fo...
     196
             i enter that gunner seat and i fear for a life
     197
                      fuck it . pic.twitter.com/Wav1bacr5j
     198
                      Fuck it. pic.twitter.com / Wav1bacr5j
     199
                    fuck it. pic.wikipedia.org / Wav1bacr5j
     [100 rows x 4 columns]
[]: df.info()
```

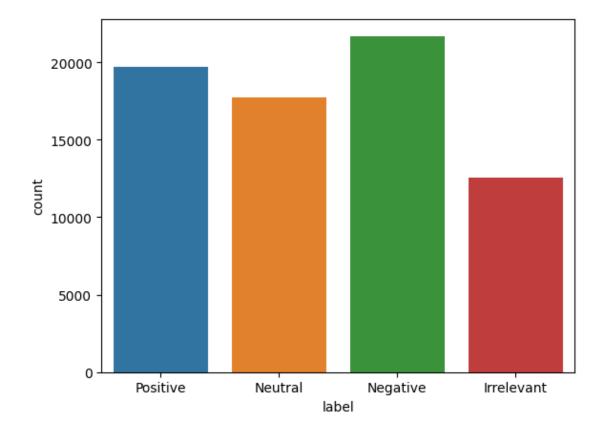
3

<class 'pandas.core.frame.DataFrame'>

```
Data columns (total 4 columns):
                  Non-Null Count Dtype
         Column
                  74681 non-null int64
         id
     0
     1
         country 74681 non-null object
         label
                  74681 non-null
                                   object
         text
                  73995 non-null
                                   object
    dtypes: int64(1), object(3)
    memory usage: 2.3+ MB
[]: df.describe(include=object)
[]:
                                      label
                          country
                                               text
     count
                            74681
                                      74681
                                             73995
                                             69490
     unique
                               32
     top
             {\tt TomClancysRainbowSix}
                                   Negative
                                      22542
     freq
                             2400
                                                172
[]: df.duplicated().sum()
[]: 2700
[]: df.drop_duplicates(inplace=True)
[]: df.duplicated().sum()
[]: 0
[]: df.isnull().sum()
[]: id
                  0
     country
                  0
     label
                  0
                326
     text
     dtype: int64
[]: df.dropna(inplace=True)
[]: df.isnull().any()
[]: id
                False
                False
     country
     label
                False
                False
     text
     dtype: bool
```

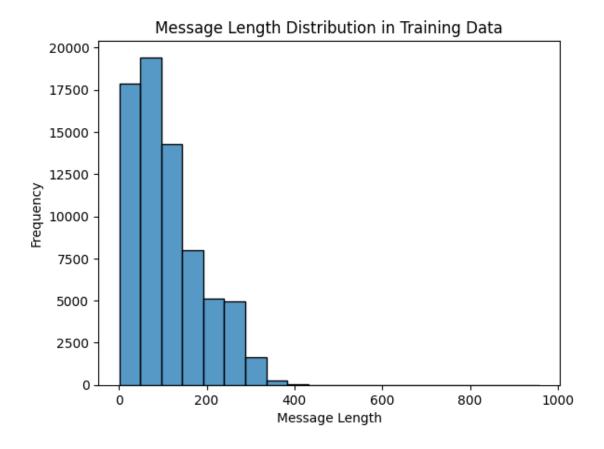
RangeIndex: 74681 entries, 0 to 74680

[]: <Axes: xlabel='label', ylabel='count'>



```
[]: message_length=(df['text']).apply(len)
    sns.histplot(x=message_length,bins=20)
    plt.title('Message Length Distribution in Training Data')
    plt.ylabel('Frequency')
    plt.xlabel('Message Length')
```

[]: Text(0.5, 0, 'Message Length')



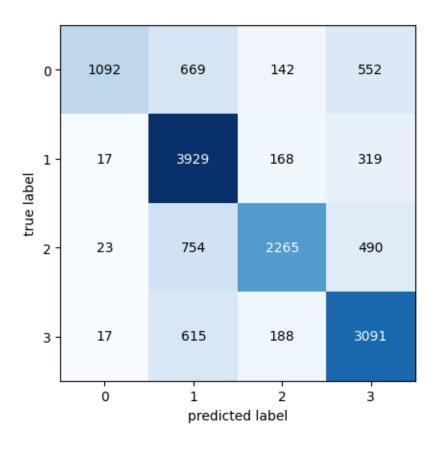
```
[]: import spacy
     # load english language model and create nlp object from it
     nlp = spacy.load("en_core_web_sm")
     # use this utility function to get the preprocessed text data
     def preprocess(text):
         # remove stop words and lemmatize the text
         doc = nlp(text)
         filtered_tokens = []
         for token in doc:
             if token.is_stop or token.is_punct:
                 continue
             filtered_tokens.append(token.lemma_)
         return " ".join(filtered_tokens)
[]: df['Preprocessed text'] = df['text'].apply(preprocess)
[]: lb=LabelEncoder()
     df['label']=lb.fit_transform(df['label'])
```

```
[]: df
[]:
              id
                       country label
     0
            2401
                  Borderlands
                                    3
     1
            2401
                  Borderlands
                                    3
                                    3
     2
            2401
                  Borderlands
     3
            2401
                  Borderlands
                                    3
            2401
                  Borderlands
                                    3
     74676 9200
                        Nvidia
                                    3
     74677
            9200
                        Nvidia
                                    3
     74678
            9200
                        Nvidia
                                    3
     74679
                        Nvidia
                                    3
            9200
     74680
            9200
                        Nvidia
                                    3
                                                            text \
     0
            I am coming to the borders and I will kill you...
     1
            im getting on borderlands and i will kill you ...
     2
            im coming on borderlands and i will murder you...
     3
            im getting on borderlands 2 and i will murder ...
     4
            im getting into borderlands and i can murder y...
     74676
            Just realized that the Windows partition of my...
     74677
            Just realized that my Mac window partition is ...
     74678
            Just realized the windows partition of my Mac ...
            Just realized between the windows partition of ...
     74679
     74680
            Just like the windows partition of my Mac is 1...
                                              Preprocessed text
     0
                                               come border kill
     1
                                         m get borderland kill
     2
                                      m come borderland murder
     3
                                     m get borderland 2 murder
     4
                                       m get borderland murder
     74676
            realize Windows partition Mac like 6 year Nvid...
     74677
            realize Mac window partition 6 year Nvidia dri...
            realize window partition Mac 6 year Nvidia dri...
     74678
     74679
            realize window partition Mac like 6 year Nvidi...
            like window partition Mac like 6 year driver i...
     74680
     [71655 rows x 5 columns]
[]: tv=TfidfVectorizer()
     df_tv=tv.fit_transform(df['Preprocessed text'])
[]: print(df_tv)
```

```
(0, 14186)
              0.5019686782389964
(0, 4300)
              0.7503332981844422
(0, 5882)
              0.43014809973153667
(1, 4303)
              0.6308352317883091
(1, 10718)
              0.4731922339217186
(1, 14186)
              0.6149276543551802
(2, 16730)
              0.7359220742014858
(2, 4303)
              0.519630312809822
(2, 5882)
              0.4340541886817236
(3, 16730)
              0.7497229075893237
(3, 4303)
              0.5293750013057333
(3, 10718)
              0.3970864765115596
(4, 16730)
              0.7497229075893237
(4, 4303)
              0.5293750013057333
(4, 10718)
              0.3970864765115596
(5, 16356)
              0.32986143201396134
(5, 5868)
              0.0950308449908003
(5, 25306)
              0.12371465037450177
(5, 18780)
              0.12279967472353039
(5, 8680)
              0.17199301599436456
(5, 6478)
              0.31519414526267836
           0.2882003846504435
(5, 26163)
(5, 12710)
             0.23515040647542382
(5, 17993)
             0.2103819690143733
(5, 18508)
           0.17463994232150065
    :
(71652, 17401)
                      0.335661757431383
(71652, 12602)
                      0.28873546946764583
(71652, 20209)
                      0.3195397101596675
(71652, 27556)
                    0.21093083092118967
(71653, 18390)
                      0.41917259340568874
(71653, 17512)
                      0.20004410985809554
(71653, 26966)
                      0.30984190903656667
(71653, 8064) 0.2857211695158495
(71653, 4956) 0.2711117868352008
(71653, 7524) 0.31837801158630585
(71653, 15399)
                      0.32982978949582387
(71653, 17401)
                      0.2933694892495072
(71653, 12602)
                      0.25235575793365683
(71653, 20209)
                      0.2792787664637086
(71653, 10264)
                      0.19437024500723696
(71653, 27556)
                      0.18435424579749274
(71653, 14875)
                      0.15320656386788417
(71654, 18390)
                      0.48735842343812535
(71654, 26966)
                      0.36024317113922943
(71654, 8064) 0.3321987670681811
(71654, 15399)
                   0.3834824335856304
(71654, 17401)
                    0.34109122116939317
```

```
(71654, 12602)
                            0.29340588165087583
      (71654, 27556)
                            0.21434272182731726
      (71654, 14875)
                            0.3562566379656403
[]: x_train, x_test, y_train, y_test = train_test_split(df_tv,_

df['label'],test_size=0.2, random_state=42)
[]: x_test.shape
[]: (14331, 28054)
[]: y_test.shape
[]: (14331,)
[]: nb=MultinomialNB()
     nb.fit(x_train,y_train)
     y_pred_nb=nb.predict(x_test)
     print('classification_report:\n',classification_report(y_test,y_pred_nb))
     print('accuracy:',accuracy_score(y_test,y_pred_nb)*100)
     print('Error value',np.mean(y_pred_nb!=y_test)*100)
     print('confusion_matrix\n',confusion_matrix(y_test,y_pred_nb))
    classification_report:
                   precision
                                recall f1-score
                                                   support
                                 0.44
               0
                       0.95
                                           0.61
                                                      2455
               1
                       0.66
                                 0.89
                                           0.76
                                                      4433
               2
                       0.82
                                 0.64
                                           0.72
                                                      3532
               3
                       0.69
                                 0.79
                                           0.74
                                                      3911
                                           0.72
                                                    14331
        accuracy
       macro avg
                       0.78
                                 0.69
                                           0.71
                                                    14331
                                 0.72
                                           0.72
                                                    14331
    weighted avg
                       0.76
    accuracy: 72.40946200544275
    Error value 27.590537994557252
    confusion matrix
     [[1092 669 142 552]
     [ 17 3929 168 319]
     [ 23 754 2265 490]
     [ 17 615 188 3091]]
[]: plot_confusion_matrix(confusion_matrix(y_test,y_pred_nb))
[]: (<Figure size 640x480 with 1 Axes>,
      <Axes: xlabel='predicted label', ylabel='true label'>)
```



```
[]: lg=LogisticRegression()
    lg.fit(x_train,y_train)
    y_pred_lg=lg.predict(x_test)
    print('classification_report:\n',classification_report(y_test,y_pred_lg))
    print('accuracy:',accuracy_score(y_test,y_pred_lg)*100)
    print('Error value',np.mean(y_pred_lg!=y_test)*100)
    print('confusion_matrix\n',confusion_matrix(y_test,y_pred_lg))
```

### classification\_report:

	precision	recall	f1-score	support
0	0.81	0.66	0.73	2455
1	0.80	0.83	0.81	4433
2	0.73	0.76	0.75	3532
3	0.75	0.79	0.77	3911
accuracy	0.10	0.10	0.77	14331
macro avg	0.77	0.76	0.76	14331
weighted avg	0.77	0.77	0.77	14331

accuracy: 77.13348684669597

```
Error value 22.866513153304027

confusion_matrix

[[1616 263 246 330]

[ 119 3666 324 324]

[ 127 352 2690 363]

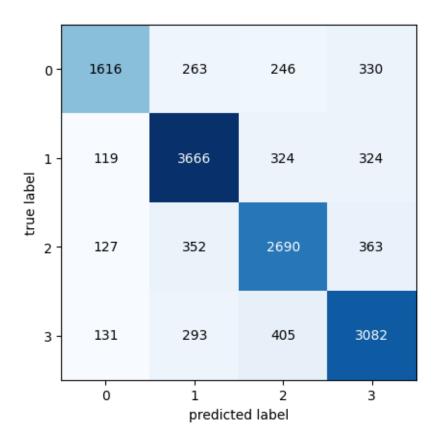
[ 131 293 405 3082]]
```

/usr/local/lib/python3.10/dist-packages/sklearn/linear\_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
Please also refer to the documentation for alternative solver options:
 https://scikit-learn.org/stable/modules/linear\_model.html#logisticregression
 n\_iter\_i = \_check\_optimize\_result(

## []: plot\_confusion\_matrix(confusion\_matrix(y\_test,y\_pred\_lg))

### 



```
[]: nltk.download('vader_lexicon')
    sid = SentimentIntensityAnalyzer()
    [nltk_data] Downloading package vader_lexicon to /root/nltk_data...
                 Package vader_lexicon is already up-to-date!
    [nltk_data]
[ ]: def get_sentiment_scores(text):
        sentiment_scores = sid.polarity_scores(text)
        return sentiment_scores
[]: df3=df.copy()
[]: df3['sentiment_scores'] = df3['text'].apply(get_sentiment_scores)
[]: df3['sentiment_scores']
             {'neg': 0.343, 'neu': 0.657, 'pos': 0.0, 'comp...
[]: 0
             {'neg': 0.37, 'neu': 0.63, 'pos': 0.0, 'compou...
    1
             {'neg': 0.37, 'neu': 0.63, 'pos': 0.0, 'compou…
    3
             {'neg': 0.343, 'neu': 0.657, 'pos': 0.0, 'comp...
             {'neg': 0.37, 'neu': 0.63, 'pos': 0.0, 'compou...
             {'neg': 0.086, 'neu': 0.817, 'pos': 0.097, 'co...
    74676
    74677
             {'neg': 0.104, 'neu': 0.896, 'pos': 0.0, 'comp...
             {'neg': 0.091, 'neu': 0.909, 'pos': 0.0, 'comp...
    74678
             {'neg': 0.074, 'neu': 0.842, 'pos': 0.084, 'co...
    74679
    74680
             {'neg': 0.09, 'neu': 0.728, 'pos': 0.182, 'com...
    Name: sentiment_scores, Length: 71655, dtype: object
[]: df3['compound'] = df3['sentiment_scores'].apply(lambda x: x['compound'])
    df3['positive'] = df3['sentiment_scores'].apply(lambda x: x['pos'])
    df3['neutral'] = df3['sentiment_scores'].apply(lambda x: x['neu'])
    df3['negative'] = df3['sentiment_scores'].apply(lambda x: x['neg'])
[]: df3['sentiment'] = df3['compound'].apply(lambda x: 'Positive' if x >= 0.05 else_
      []: df3
[]:
             id
                     country label
                                    \
    0
           2401 Borderlands
                                  3
           2401 Borderlands
    1
                                  3
           2401 Borderlands
    2
                                  3
    3
           2401 Borderlands
                                  3
    4
           2401 Borderlands
                                  3
    74676 9200
                      Nvidia
                                  3
```

```
74677
       9200
                   Nvidia
                               3
                               3
74678
       9200
                   Nvidia
74679
       9200
                   Nvidia
                               3
74680
       9200
                   Nvidia
                               3
                                                       text \
0
       I am coming to the borders and I will kill you...
1
       im getting on borderlands and i will kill you ...
2
       im coming on borderlands and i will murder you...
3
       im getting on borderlands 2 and i will murder ...
4
       im getting into borderlands and i can murder y...
74676
       Just realized that the Windows partition of my...
74677
       Just realized that my Mac window partition is ...
74678
       Just realized the windows partition of my Mac ...
74679
       Just realized between the windows partition of...
74680
       Just like the windows partition of my Mac is 1...
                                         Preprocessed text
0
                                          come border kill
1
                                     m get borderland kill
2
                                 m come borderland murder
3
                                m get borderland 2 murder
4
                                   m get borderland murder
74676
       realize Windows partition Mac like 6 year Nvid...
74677
       realize Mac window partition 6 year Nvidia dri...
       realize window partition Mac 6 year Nvidia dri...
74678
74679
       realize window partition Mac like 6 year Nvidi...
74680
       like window partition Mac like 6 year driver i...
                                                             compound
                                                                       positive \
                                          sentiment_scores
0
       {'neg': 0.343, 'neu': 0.657, 'pos': 0.0, 'comp...
                                                            -0.6908
                                                                         0.000
       {'neg': 0.37, 'neu': 0.63, 'pos': 0.0, 'compou...
1
                                                            -0.6908
                                                                         0.000
2
       {'neg': 0.37, 'neu': 0.63, 'pos': 0.0, 'compou...
                                                            -0.6908
                                                                         0.000
3
       {'neg': 0.343, 'neu': 0.657, 'pos': 0.0, 'comp...
                                                            -0.6908
                                                                         0.000
4
       {'neg': 0.37, 'neu': 0.63, 'pos': 0.0, 'compou...
                                                            -0.6908
                                                                         0.000
       {'neg': 0.086, 'neu': 0.817, 'pos': 0.097, 'co...
74676
                                                             0.0772
                                                                         0.097
       {'neg': 0.104, 'neu': 0.896, 'pos': 0.0, 'comp...
74677
                                                            -0.2960
                                                                         0.000
       {'neg': 0.091, 'neu': 0.909, 'pos': 0.0, 'comp...
74678
                                                            -0.2960
                                                                         0.000
74679
       {'neg': 0.074, 'neu': 0.842, 'pos': 0.084, 'co...
                                                             0.0772
                                                                         0.084
74680
       {'neg': 0.09, 'neu': 0.728, 'pos': 0.182, 'com...
                                                             0.3687
                                                                         0.182
       neutral
                negative sentiment
0
         0.657
                    0.343
                           Negative
1
         0.630
                    0.370
                           Negative
```

```
2
         0.630
                  0.370 Negative
3
         0.657
                  0.343
                         Negative
4
         0.630
                  0.370
                         Negative
         0.817
74676
                  0.086 Positive
                  0.104 Negative
74677
         0.896
74678
         0.909
                  0.091
                         Negative
74679
         0.842
                  0.074 Positive
74680
         0.728
                  0.090 Positive
```

[71655 rows x 11 columns]

### []: df3['sentiment']

```
Negative
[]: 0
              Negative
     1
     2
              Negative
     3
              Negative
              Negative
     4
     74676
              Positive
     74677
              Negative
     74678
              Negative
     74679
              Positive
     74680
              Positive
     Name: sentiment, Length: 71655, dtype: object
```