# vips-project

### November 9, 2024

```
[]: import pandas as pd
     import numpy as np
     import re
     import nltk
     import seaborn as sns
     import matplotlib.pyplot as plt
     from matplotlib import style
     style.use('ggplot')
     from textblob import TextBlob
     from nltk.tokenize import word_tokenize
     from nltk.stem import PorterStemmer
     from nltk.corpus import stopwords
     nltk.download('stopwords')
     stop_words = set(stopwords.words('english'))
     from wordcloud import WordCloud
     from sklearn.feature_extraction.text import CountVectorizer
     from sklearn.model_selection import train_test_split
     from sklearn.linear_model import LogisticRegression
     from sklearn.metrics import accuracy_score, classification_report,_
      ⇔confusion_matrix, ConfusionMatrixDisplay
    [nltk_data] Downloading package stopwords to /root/nltk_data...
    [nltk data]
                  Unzipping corpora/stopwords.zip.
[]: df = pd.read_csv('/content/vaccination_tweets.csv')
[]: df.head()
[]:
                                                               user_location \
                                        user_name
     0 1340539111971516416
                                       Rachel Roh La Crescenta-Montrose, CA
     1 1338158543359250433
                                      Albert Fong
                                                           San Francisco, CA
     2 1337858199140118533
                                         eli
                                                                  Your Bed
     3 1337855739918835717
                                    Charles Adler
                                                      Vancouver, BC - Canada
     4 1337854064604966912 Citizen News Channel
                                                                         NaN
                                         user_description
                                                                  user_created \
     O Aggregator of Asian American news; scanning di... 2009-04-08 17:52:46
     1 Marketing dude, tech geek, heavy metal & '80s ... 2009-09-21 15:27:30
```

```
2
                                       heil, hydra
                                                       2020-06-25 23:30:28
3 Hosting "CharlesAdlerTonight" Global News Radi...
                                                      2008-09-10 11:28:53
4 Citizen News Channel bringing you an alternati...
                                                      2020-04-23 17:58:42
   user_followers
                   user_friends
                                 user_favourites
                                                   user_verified
0
              405
                            1692
                                              3247
                                                            False
              834
                             666
                                               178
                                                            False
1
2
               10
                              88
                                               155
                                                            False
3
                            3933
            49165
                                             21853
                                                             True
4
                             580
                                                            False
              152
                                              1473
                  date
   2020-12-20 06:06:44
                        Same folks said daikon paste could treat a cyt...
1 2020-12-13 16:27:13
                         While the world has been on the wrong side of ...
2 2020-12-12 20:33:45
                         #coronavirus #SputnikV #AstraZeneca #PfizerBio...
                         Facts are immutable, Senator, even when you're...
3 2020-12-12 20:23:59
4 2020-12-12 20:17:19
                         Explain to me again why we need a vaccine @Bor...
                                              hashtags
                                                                      source
0
                                   ['PfizerBioNTech']
                                                        Twitter for Android
1
                                                   NaN
                                                            Twitter Web App
   ['coronavirus', 'SputnikV', 'AstraZeneca', 'Pf... Twitter for Android
3
                                                            Twitter Web App
                                                   NaN
      ['whereareallthesickpeople', 'PfizerBioNTech']
                                                         Twitter for iPhone
   retweets
            favorites
                         is retweet
                              False
0
1
          1
                      1
                              False
2
          0
                      0
                              False
                  2129
3
        446
                              False
4
                              False
          0
                      0
```

#### []: df.info()

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 11020 entries, 0 to 11019
Data columns (total 16 columns):

#	Column	Non-Null Count	Dtype
0	id	11020 non-null	int64
1	user_name	11020 non-null	object
2	${\tt user\_location}$	8750 non-null	object
3	${\tt user\_description}$	10340 non-null	object
4	user_created	11020 non-null	object
5	user_followers	11020 non-null	int64
6	user_friends	11020 non-null	int64
7	user_favourites	11020 non-null	int64

```
user_verified
               9
                                                                              11020 non-null object
                          date
               10 text
                                                                              11020 non-null
                                                                                                                           object
               11 hashtags
                                                                              8438 non-null
                                                                                                                             object
               12
                        source
                                                                              11019 non-null
                                                                                                                            object
               13 retweets
                                                                              11020 non-null
                                                                                                                            int64
               14 favorites
                                                                              11020 non-null
                                                                                                                          int64
                                                                              11020 non-null bool
               15 is retweet
            dtypes: bool(2), int64(6), object(8)
            memory usage: 1.2+ MB
[]: df.isnull().sum()
[]: id
                                                                                0
                                                                                0
             user_name
             user_location
                                                                       2270
             user_description
                                                                          680
             user_created
                                                                                0
             user_followers
                                                                                0
             user_friends
                                                                                0
                                                                                0
             user_favourites
             user_verified
                                                                                0
                                                                                0
             date
             text
                                                                                0
             hashtags
                                                                       2582
              source
                                                                                1
             retweets
                                                                                0
             favorites
                                                                                0
              is_retweet
                                                                                0
              dtype: int64
[]: df.columns
[]: Index(['id', 'user_name', 'user_location', 'user_description', 'user_created',
                                   'user_followers', 'user_friends', 'user_favourites', 'user_verified',
                                   'date', 'text', 'hashtags', 'source', 'retweets', 'favorites',
                                   'is retweet'],
                               dtype='object')
[]: text_df = df.drop(['id', 'user_name', 'user_location', 'user_description', user_location', 'user_description', user_location', user_locati
                 'user_followers', 'user_friends', 'user_favourites', 'user_verified',
                                   'date', 'hashtags', 'source', 'retweets', 'favorites',
                                   'is_retweet'], axis=1)
              text_df.head()
```

11020 non-null

bool

8

[]: text

- O Same folks said daikon paste could treat a cyt...
- 1 While the world has been on the wrong side of ...
- 2 #coronavirus #SputnikV #AstraZeneca #PfizerBio...
- 3 Facts are immutable, Senator, even when you're...
- 4 Explain to me again why we need a vaccine @Bor...

```
[]: print(text_df['text'].iloc[0],"\n")
    print(text_df['text'].iloc[1],"\n")
    print(text_df['text'].iloc[2],"\n")
    print(text_df['text'].iloc[3],"\n")
    print(text_df['text'].iloc[4],"\n")
```

Same folks said daikon paste could treat a cytokine storm #PfizerBioNTech https://t.co/xeHhIMg1kF

While the world has been on the wrong side of history this year, hopefully, the biggest vaccination effort we've ev... https://t.co/dlCHrZjkhm

#coronavirus #SputnikV #AstraZeneca #PfizerBioNTech #Moderna #Covid\_19 Russian
vaccine is created to last 2-4 years... https://t.co/ieYlCKBr8P

Facts are immutable, Senator, even when you're not ethically sturdy enough to acknowledge them. (1) You were born i.m. https://t.co/jqgV18kch4

Explain to me again why we need a vaccine @BorisJohnson @MattHancock #whereareallthesickpeople #PfizerBioNTech... https://t.co/KxbSRoBEHq

```
[]: def data_processing(text):
    text = text.lower()
    text = re.sub(r"https\S+|www\S+https\S+", '',text, flags=re.MULTILINE)
    text = re.sub(r'\@w+|\#','',text)
    text = re.sub(r'[^\w\s]','',text)
    text_tokens = word_tokenize(text)
    filtered_text = [w for w in text_tokens if not w in stop_words]
    return " ".join(filtered_text)
```

```
[]: !pip install nltk
  import nltk
  import re
  from nltk.corpus import stopwords
  from nltk.tokenize import word_tokenize

  nltk.download('punkt') # download punkt resource

stop_words = set(stopwords.words('english'))
```

```
text_df.text = text_df['text'].apply(data_processing)
    Requirement already satisfied: nltk in /usr/local/lib/python3.10/dist-packages
    Requirement already satisfied: click in /usr/local/lib/python3.10/dist-packages
    (from nltk) (8.1.7)
    Requirement already satisfied: joblib in /usr/local/lib/python3.10/dist-packages
    (from nltk) (1.4.2)
    Requirement already satisfied: regex>=2021.8.3 in
    /usr/local/lib/python3.10/dist-packages (from nltk) (2024.9.11)
    Requirement already satisfied: tqdm in /usr/local/lib/python3.10/dist-packages
    (from nltk) (4.66.6)
    [nltk_data] Downloading package punkt to /root/nltk_data...
                  Unzipping tokenizers/punkt.zip.
    [nltk data]
[]: text_df = text_df.drop_duplicates('text')
[]: stemmer = PorterStemmer()
     def stemming(data):
         text = [stemmer.stem(word) for word in data]
         return data
[]: text_df['text'].apply(lambda x: stemming(x))
[]: 0
              folks said daikon paste could treat cytokine s...
     1
              world wrong side history year hopefully bigges...
     2
              coronavirus sputnikv astrazeneca pfizerbiontec...
     3
              facts immutable senator even youre ethically s...
              explain need vaccine borisjohnson matthancock ...
     11014
              ruvanw gmalavige chandi2012 already 3rd pfizer...
     11015
              number covid19 cases today 17 november 2021 re...
     11017
              number covid19 cases today 16 november 2021 re...
     11018
              pfizerbiontech developed new pill paxlovid inc...
     11019
              number covid19 cases today 15 november 2021 re...
     Name: text, Length: 10543, dtype: object
[ ]: text_df.head()
[]:
                                                      text
     O folks said daikon paste could treat cytokine s...
     1 world wrong side history year hopefully bigges...
     2 coronavirus sputnikv astrazeneca pfizerbiontec...
     3 facts immutable senator even youre ethically s...
     4 explain need vaccine borisjohnson matthancock ...
```

```
[]: print(text_df['text'].iloc[0],"\n")
    print(text_df['text'].iloc[1],"\n")
    print(text_df['text'].iloc[2],"\n")
    print(text_df['text'].iloc[3],"\n")
    print(text_df['text'].iloc[4],"\n")
    folks said daikon paste could treat cytokine storm pfizerbiontech
    world wrong side history year hopefully biggest vaccination effort weve ev
    coronavirus sputnikv astrazeneca pfizerbiontech moderna covid_19 russian vaccine
    created last 24 years
    facts immutable senator even youre ethically sturdy enough acknowledge 1 born
    explain need vaccine borisjohnson matthancock whereareallthesickpeople
    pfizerbiontech
[]: text_df.info()
    <class 'pandas.core.frame.DataFrame'>
    Index: 10543 entries, 0 to 11019
    Data columns (total 1 columns):
         Column Non-Null Count Dtype
    --- ----- ------
         text
                 10543 non-null object
    dtypes: object(1)
    memory usage: 164.7+ KB
[]: def polarity(text):
        return TextBlob(text).sentiment.polarity
    text_df['polarity'] = text_df['text'].apply(polarity)
[]: text_df.head(10)
[]:
                                                    text
                                                          polarity
    O folks said daikon paste could treat cytokine s...
                                                           0.000
    1 world wrong side history year hopefully bigges...
                                                           -0.500
    2 coronavirus sputnikv astrazeneca pfizerbiontec...
                                                           0.000
    3 facts immutable senator even youre ethically s...
                                                           0.100
    4 explain need vaccine borisjohnson matthancock ...
                                                           0.000
    5 anyone useful adviceguidance whether covid vac...
                                                           0.400
```

-0.100

0.675

0.000

0.000

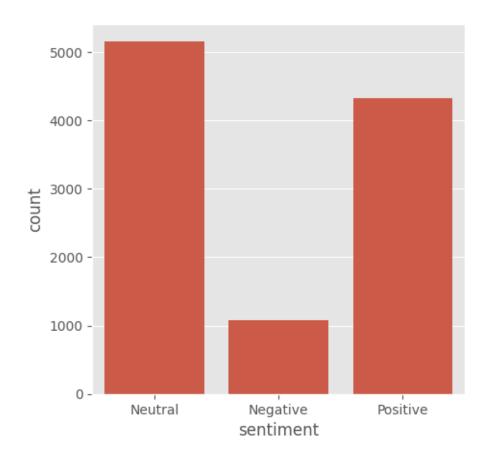
6 bit sad claim fame success vaccination patriot...

7 many bright days 2020 best 1 bidenharris winni...

8 covid vaccine getting covidvaccine covid19 pfi...

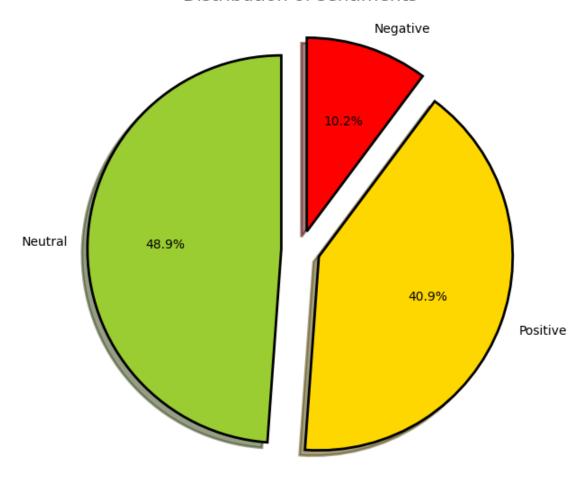
9 covidvaccine states start getting covid19vacci...

```
[]: def sentiment(label):
         if label < 0:</pre>
             return "Negative"
         elif label == 0:
            return "Neutral" # This line was missing, causing the IndentationError
         elif label > 0:
            return "Positive"
[]: text_df['sentiment'] = text_df['polarity'].apply(sentiment)
[]: text_df.head()
[]:
                                                     text polarity sentiment
    0 folks said daikon paste could treat cytokine s...
                                                              0.0
                                                                    Neutral
     1 world wrong side history year hopefully bigges...
                                                             -0.5 Negative
     2 coronavirus sputnikv astrazeneca pfizerbiontec...
                                                              0.0
                                                                   Neutral
     3 facts immutable senator even youre ethically s...
                                                              0.1 Positive
     4 explain need vaccine borisjohnson matthancock ...
                                                              0.0 Neutral
[]: fig = plt.figure(figsize=(5,5))
     sns.countplot(x='sentiment', data=text_df)
[]: <Axes: xlabel='sentiment', ylabel='count'>
```



[]: Text(0.5, 1.0, 'Distribution\xa0of\xa0sentiments')

# Distribution of sentiments

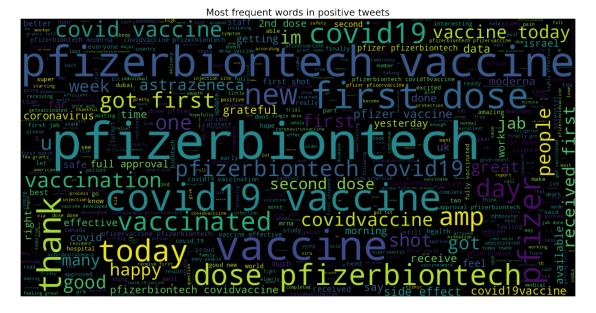


```
[]: pos_tweets = text_df[text_df.sentiment == 'Positive']
  pos_tweets = pos_tweets.sort_values(['polarity'], ascending= False)
  pos_tweets.head()
```

[]: text polarity sentiment
9317 best way get merrygoround pfizer pfizerbiontec... 1.0 Positive
2340 applying emotion pfizerbiontech based best evi... 1.0 Positive
6295 pfizer jab morning efficient wellorganised tha... 1.0 Positive
5041 get art printed awesome products support redbu... 1.0 Positive
1055 already vaccinated getting vaccine soon plan t... 1.0 Positive

```
[]: pos_tweets = text_df[text_df.sentiment == 'Positive']
  pos_tweets = pos_tweets.sort_values(['polarity'], ascending= False)
  pos_tweets.head()
```

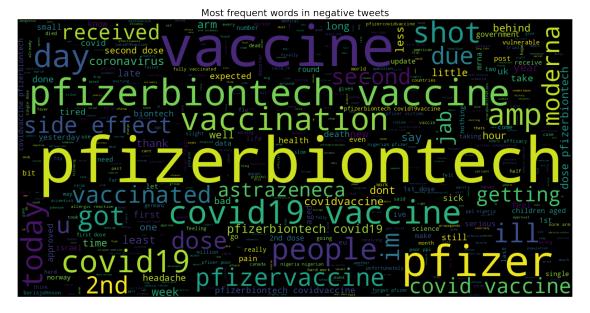
```
[]:
                                                        text polarity sentiment
    9317 best way get merrygoround pfizer pfizerbiontec...
                                                                 1.0 Positive
     2340 applying emotion pfizerbiontech based best evi...
                                                                 1.0 Positive
     6295 pfizer jab morning efficient wellorganised tha...
                                                                 1.0 Positive
     5041 get art printed awesome products support redbu...
                                                                 1.0 Positive
     1055 already vaccinated getting vaccine soon plan t...
                                                                 1.0 Positive
[]: text = ' '.join([word for word in pos_tweets['text']])
     plt.figure(figsize=(20,15), facecolor='None')
     wordcloud = WordCloud(max_words=500, width=1600, height=800).generate(text)
     plt.imshow(wordcloud, interpolation='bilinear')
     plt.axis("off")
     plt.title('Most frequent words in positive tweets', fontsize=19)
     plt.show()
```



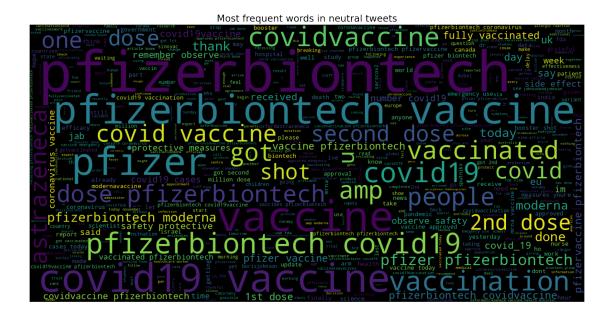
```
[]: neg_tweets = text_df[text_df.sentiment == 'Negative']
neg_tweets = neg_tweets.sort_values(['polarity'], ascending= False)
neg_tweets.head()

[]: text polarity sentiment
2912 work skilled nursing facility got first vaccin... -0.003333 Negative
7256 200321 752308 vaccinations new daily record da... -0.003409 Negative
2073 ukgovernment cant even vaccinate properly ethi... -0.004762 Negative
7715 got first dose less waiting time airport vacci... -0.005556 Negative
7157 nas_k27 second dose due end next month well fa... -0.006250 Negative
```

```
[]: text = ' '.join([word for word in neg_tweets['text']])
    plt.figure(figsize=(20,15), facecolor='None')
    wordcloud = WordCloud(max_words=500, width=1600, height=800).generate(text)
    plt.imshow(wordcloud, interpolation='bilinear')
    plt.axis("off")
    plt.title('Most frequent words in negative tweets', fontsize=19)
    plt.show()
```



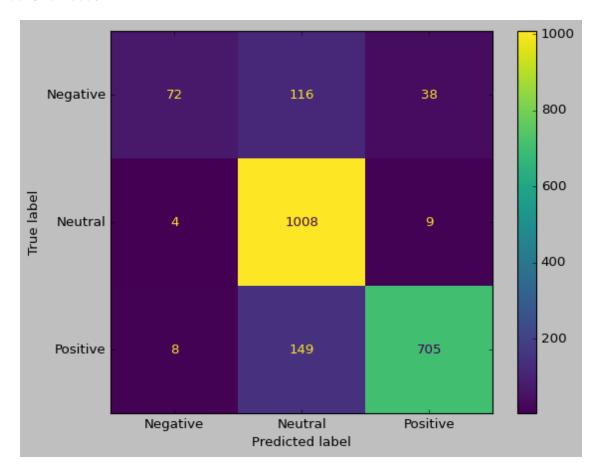
```
[]: neutral_tweets = text_df[text_df.sentiment == 'Neutral']
     neutral_tweets = neutral_tweets.sort_values(['polarity'], ascending= False)
     neutral_tweets.head()
[]:
                                                        text polarity sentiment
           folks said daikon paste could treat cytokine s...
                                                                 0.0
                                                                        Neutral
     7347 anyone else feel like framing vaccine card pfi...
                                                                 0.0
                                                                        Neutral
     7458 looking forward getting second pfizer shot any...
                                                                 0.0
                                                                        Neutral
     7454 never thought id running diff vaccine modernav...
                                                                 0.0
                                                                        Neutral
     7453
          john__m dont get choose one person know asked...
                                                                 0.0
                                                                        Neutral
[]: text = ' '.join([word for word in neutral_tweets['text']])
     plt.figure(figsize=(20,15), facecolor='None')
     wordcloud = WordCloud(max_words=500, width=1600, height=800).generate(text)
     plt.imshow(wordcloud, interpolation='bilinear')
     plt.axis("off")
     plt.title('Most frequent words in neutral tweets', fontsize=19)
     plt.show()
```



```
[]: vect = CountVectorizer(ngram range=(1,2)).fit(text_df['text'])
[]: from sklearn.feature_extraction.text import CountVectorizer
     vect = CountVectorizer(ngram_range=(1,2)).fit(text_df['text'])
     # Replace get feature names with get feature names out
     feature_names = vect.get_feature_names_out()
     print("Number of features: {}\n".format(len(feature_names)))
     print("First 20 features:\n {}".format(feature_names[:20]))
    Number of features: 78583
    First 20 features:
     ['000' '000 doses' '000 initial' '000 people' '000 vaccines' '0000001'
     '0000001 covid19' '0011' '0011 abt' '004' '004 covid' '004 israelis' '01'
     '01 getting' '01 june' '01 november' '01aug2021' '01aug2021 doublevaxxed'
     '02' '02 175']
[]: X = text_df['text']
     Y = text df['sentiment']
     X = vect.transform(X)
[]: from sklearn.model_selection import train_test_split
     # Assuming X and Y are your features and target variable respectively
     # Split the data into training and testing sets
```

```
x_train, x_test, y_train, y_test = train_test_split(X, Y, test_size=0.2,_
      ⊖random_state=42) # Adjust test_size and random_state as needed
     print("Size of x_train:", (x_train.shape))
     print("Size of x_train:", (x_train.shape))
     print("Size of y train:", (y train.shape))
     print("Size of x_test:", (x_test.shape))
     print("Size of y_test:", (y_test.shape))
    Size of x train: (8434, 78583)
    Size of x_train: (8434, 78583)
    Size of y_train: (8434,)
    Size of x_test: (2109, 78583)
    Size of y_test: (2109,)
[]: import warnings
     warnings.filterwarnings('ignore')
[]: logreg = LogisticRegression()
     logreg.fit(x_train, y_train)
     logreg_pred = logreg.predict(x_test)
     logreg_acc = accuracy_score(logreg_pred, y_test)
     print("Test accuracy: {:.2f}%".format(logreg_acc*100))
    Test accuracy: 84.64%
[]: print(confusion_matrix(y_test, logreg_pred))
     print("\n")
     print(classification_report(y_test, logreg_pred))
    [[ 72 116
                  38]
         4 1008
                   9]
                 705]]
         8 149
                               recall f1-score
                  precision
                                                   support
                                 0.32
                                           0.46
                                                       226
        Negative
                       0.86
                                 0.99
         Neutral
                       0.79
                                            0.88
                                                      1021
        Positive
                       0.94
                                 0.82
                                           0.87
                                                       862
                                           0.85
                                                      2109
        accuracy
                                            0.74
                                                      2109
       macro avg
                       0.86
                                 0.71
    weighted avg
                       0.86
                                 0.85
                                           0.83
                                                      2109
```

[]: <sklearn.metrics.\_plot.confusion\_matrix.ConfusionMatrixDisplay at 0x7a23bc41aaa0>



```
Best parameters: {'C': 10}
[]: y_pred = grid.predict(x_test)
[]: logreg_acc = accuracy_score(y_pred, y_test)
     print("Test accuracy: {:.2f}%".format(logreg_acc*100))
    Test accuracy: 86.30%
[]: print(confusion_matrix(y_test, y_pred))
     print("\n")
     print(classification_report(y_test, y_pred))
    [[ 90 101
                  351
         5 1006
                  10]
         9 129 724]]
                  precision
                               recall f1-score
                                                   support
                                 0.40
        Negative
                       0.87
                                            0.55
                                                       226
         Neutral
                       0.81
                                 0.99
                                            0.89
                                                      1021
        Positive
                       0.94
                                 0.84
                                            0.89
                                                       862
                                                      2109
        accuracy
                                            0.86
       macro avg
                       0.87
                                 0.74
                                            0.77
                                                      2109
    weighted avg
                       0.87
                                 0.86
                                            0.85
                                                      2109
[]: from sklearn.svm import LinearSVC
[]: SVCmodel = LinearSVC()
     SVCmodel.fit(x_train, y_train)
[]: LinearSVC()
[]: svc_pred = SVCmodel.predict(x_test)
     svc_acc = accuracy_score(svc_pred, y_test)
     print("test accuracy: {:.2f}%".format(svc_acc*100))
    test accuracy: 87.34%
[]: print(confusion_matrix(y_test, svc_pred))
     print("\n")
     print(classification_report(y_test, svc_pred))
    [[ 101
             91
                  34]
         6 1007
     Γ
                   81
```

## [ 14 114 734]]

precision

```
Negative
                       0.83
                                 0.45
                                           0.58
                                                       226
         Neutral
                       0.83
                                 0.99
                                            0.90
                                                      1021
        Positive
                       0.95
                                 0.85
                                           0.90
                                                       862
                                           0.87
                                                      2109
        accuracy
                       0.87
                                 0.76
                                            0.79
                                                      2109
       macro avg
    weighted avg
                       0.88
                                 0.87
                                           0.87
                                                      2109
[]: from sklearn.model_selection import GridSearchCV
     from sklearn.svm import LinearSVC
     SVCmodel = LinearSVC()
     SVCmodel.fit(x_train, y_train)
     # Remove 'kernel', 'degree', and 'gamma' from the grid
     grid = {
         'C': [0.01, 0.1, 1, 10],
         # 'kernel': ["linear", "poly", "rbf", "sigmoid"], # Remove this line
         # 'degree': [1, 3, 5, 7], # Remove this line
         # 'qamma': [0.01, 1] # Remove this line
     }
     grid = GridSearchCV(SVCmodel, param_grid=grid)
     grid.fit(x_train, y_train)
[]: GridSearchCV(estimator=LinearSVC(), param_grid={'C': [0.01, 0.1, 1, 10]})
[]: print("Best parameter:", grid.best_params_)
    Best parameter: {'C': 10}
[]: y_pred = grid.predict(x_test)
[]: logreg_acc = accuracy_score(y_pred, y_test)
     print("Test accuracy: {:.2f}%".format(logreg_acc*100))
    Test accuracy: 87.58%
[]: print(confusion_matrix(y_test, y_pred))
     print("\n")
```

recall f1-score

support

print(classification\_report(y\_test, y\_pred))

```
[[ 105 87 34]
[ 7 1005 9]
[ 14 111 737]]
```

	precision	recall	f1-score	support
Negative	0.83	0.46	0.60	226
Neutral	0.84	0.98	0.90	1021
Positive	0.94	0.85	0.90	862
accuracy			0.88	2109
macro avg	0.87	0.77	0.80	2109
weighted avg	0.88	0.88	0.87	2109

```
[1]: import tweepy #to access the twitter api
import pandas as pd #for basic data operations
```

```
[7]: # Importing the keys from twitter api
    consumerKey = "xxxxxxxxxxxxxxxxxxx"
    consumerSecret = "xxxxxxxxxxxxxxxxxx"
    accessToken = "xxxxxxxxxxxxxxxxxxx"
    accessTokenSecret = "xxxxxxxxxxxxxxxxx"
    # Establish the connection with twitter API
    auth = tweepy.OAuthHandler(consumerKey, consumerSecret)
    auth.set_access_token(accessToken, accessTokenSecret)
    api = tweepy.API(auth)
    # Search for the Term and define number of tweets
    searchTerm = input("Enter Keyword/Tag to search about: ")
    NoOfTerms = int(input("Enter how many tweets to search: "))
    # Get no of tweets and searched term together
```

Enter Keyword/Tag to search about: covidvaccine Enter how many tweets to search: 9

[]: