import numpy as np
import pandas as pd

df=pd.read_csv('_/content/stress.csv')
df.head()

	subreddit	post_id	sentence_range	text	id	label	confidence	social_timestamp s	soc
0	ptsd	8601tu	(15, 20)	He said he had not felt that way before, sugge	33181	1	0.8	1521614353	
1	assistance	8lbrx9	(0, 5)	Hey there r/assistance, Not sure if this is th	2606	0	1.0	1527009817	
2	ptsd	9ch1zh	(15, 20)	My mom then hit me with the newspaper and it s	38816	1	0.8	1535935605	
3	relationships	7rorpp	[5, 10]	until i met my new boyfriend, he is amazing, h	239	1	0.6	1516429555	
4	survivorsofabuse	9p2gbc	[0, 5]	October is Domestic Violence Awareness Month a	1421	1	0.8	1539809005	

5 rows × 116 columns



df.describe()

	id	label	confidence	$social_timestamp$	social_karma	syntax_ari	lex_liwc_WC
count	2185.000000	2185.000000	2185.000000	2.185000e+03	2185.000000	2185.000000	2185.000000
mean	13970.466362	0.518078	0.806893	1.517997e+09	18.452174	4.700834	86.035240
std	17509.350310	0.499787	0.176911	1.546128e+07	78.752148	3.315674	32.883955
min	4.000000	0.000000	0.500000	1.483274e+09	0.000000	-6.620000	5.000000
25%	951.000000	0.000000	0.600000	1.510162e+09	2.000000	2.495056	65.000000
50%	1940.000000	1.000000	0.800000	1.516987e+09	5.000000	4.309623	81.000000
75%	25827.000000	1.000000	1.000000	1.530755e+09	11.000000	6.481267	101.000000
max	55757.000000	1.000000	1.000000	1.542592e+09	1435.000000	24.074231	310.000000
•	440						

8 rows × 112 columns



df.isnull().sum()

subreddit post_id sentence_range text id	0 0 0 0
	• •
<pre>lex_dal_avg_pleasantness</pre>	1
social_upvote_ratio	1
social_num_comments	1
syntax_fk_grade	1
sentiment	1
Length: 116, dtype: int64	

```
import nltk
import re
from nltk. corpus import stopwords
import string
nltk. download( 'stopwords' )
stemmer = nltk. SnowballStemmer("english")
stopword=set (stopwords . words ( 'english' ))
def clean(text):
    text = str(text) . lower() #returns a string where all characters are lower case. Symbols and Numbers are ignored. text = re. sub('\[.*?\]','',text) #substring and returns a string with replaced values.
    text = re. sub('https?://\S+/www\. \S+', ' ', text)#whitespace char with pattern
    text = re. sub('<. *?>+', ' ', text)#special char enclosed in square brackets
    text = re. sub(' [%s]' % re. escape(string. punctuation), ' ', text)#eliminate punctuation from string text = re. sub(' \n',' ', text)#word character ASCII punctuation
    text = [word for word in text. split(' ') if word not in stopword] #removing stopwords
    text =" ". join(text)
    text = [stemmer . stem(word) for word in text. split(' ') ]#remove morphological affixes from words
    text = " ". join(text)
    return text
df [ "text"] = df["text"]. apply(clean)
     [nltk_data] Downloading package stopwords to /root/nltk_data...
     [nltk_data] Unzipping corpora/stopwords.zip.
import matplotlib. pyplot as plt
from wordcloud import WordCloud, STOPWORDS, ImageColorGenerator
text = " ". join(i for i in df. text)
stopwords = set (STOPWORDS)
wordcloud = WordCloud( stopwords=stopwords,background_color="white") . generate(text)
plt. figure(figsize=(10, 10) )
plt. imshow(wordcloud )
plt. axis("off")
plt. show( )
                                   *see
            stillstart®
from sklearn. feature_extraction. text import CountVectorizer
from sklearn. model_selection import train_test_split
x = np.array (df["text"])
y = np.array (df["label"])
cv = CountVectorizer ()
X = cv. fit_transform(x)
print(X)
xtrain, xtest, ytrain, ytest = train_test_split(X, y,test_size=0.33)
       (0, 6558)
                      1
       (0, 2894)
                      1
       (0, 8381)
                      1
       (0, 758)
                      1
       (0, 7415)
                      1
       (0, 3315)
                      1
       (0, 6388)
       (0, 7915)
                      1
       (0, 253)
       (0, 8650)
       (0, 3812)
                      1
       (0, 4459)
                      1
       (0, 4714)
                      1
       (0, 1929)
                      1
       (0, 4534)
                      1
        (0, 2882)
                      1
       (0, 2291)
                      3
       (0, 3711)
                      1
       (0, 4706)
```

```
(0, 3269)
       (0, 7396)
       (0, 6078)
       (0, 3675)
       (0, 4581)
                     1
       (0, 1610)
                     1
       (2184, 2334) 1
       (2184, 7792) 1
       (2184, 4256)
       (2184, 7715)
       (2184, 3617)
       (2184, 652)
       (2184, 5096)
       (2184, 1888)
       (2184, 410)
       (2184, 1622) 1
       (2184, 6922)
(2184, 1611)
       (2184, 2500)
       (2184, 8514)
       (2184, 4564) 1
       (2184, 5308)
       (2184, 6555) 1
       (2184, 780)
       (2184, 1064) 1
       (2184, 402) 1
(2184, 7057) 1
       (2184, 8582) 1
       (2184, 5898) 1
       (2184, 7046) 1
       (2184, 5251) 1
from sklearn.naive_bayes import BernoulliNB
model=BernoulliNB()
model.fit(xtrain,ytrain)
     BernoulliNB()
user=input("Enter the text")
data=cv.transform([user]).toarray()
output=model.predict(data)
print(output)
     Enter the texti am feeling happy
     [1]
```

✓ 5s completed at 06:39

Literature Survey:

Sr. No	Title of	Name of	Published	Remarks
	Paper	Authors	Year	
1	Automatic Stress Detection Using Wearable Sensors and Machine Learning	Shruti Gedam, Sanchita Paul	July 1-3, 2020	In our day to day life there are many stressors like Physical, Psychological, Environmental, and Psychosocial Detecting the stress we can use many technology in this project we will be use Wearable Sensors And for extracted the features using various machine learning algorithms features extracted using Heart rate, Heart rate skin conductance are more useful in prediction of stress level Random forest and K-Nearest Neighbor this algorithms are used
2	A Decision Tree Optimised SVM Model for Stress Detection using Biosignals	Alana Paul Cruz, Aravind Pradeep, Kavali Riya Sivasankar and Krishnaveni K.S	2020	In this project for stress detection SVM model is use Linear 88.9% Quadratic 92.6% Cubic 96.3% this is a type of svm model. Following component are used heart rate,pr interval, OR interval, Qt interval, ST segment In our work we selected ECG as the bio signal and extracted its features. EDR (ECG Derived Respiration) feature can be easily derived without any extra sensors. Among those unique features we chose ECG derived Respiration, Respiration Rate,
3	Stress Detection with Machine Learning and Deep Learning using Multimodal Physiological Data	Pramod Bobade, Vani M	2020	stress detection on individuals using multimodal dataset recorded from wearable physiological and motion sensors, which can prevent a person from various stress@related health problems. Data of sensor modalities like three@axis acceleration (ACC), electrocardiogram (ECG), blood volume pulse (BVP), body temperature (TEMP), respiration (RESP), electromyogram (EMG) and electrodermal activity (EDA) are for three physiological conditions
4	Machine Learning and IoT for Prediction and Detection of Stress	Mr. Purnendu Shekhar Pandey	2017	We applied SVM and Logistic Regression which show considerable improvement over VF - 15 and Naive Bayes without any external weights provided. In the weighted VF - 15 the accuracy was comparable to SVM Logistic Regression 100 % 66 % SVM 97 % 68 %. Components Used are Node MCU & Pulse Sensor