Stress, anxiety, and depression are threatening the mental health of people. Every person has a reason for having a stressful life. People often share their feelings on social media platforms like on Instagram in the form of posts and stories, and on Reddit in the form of asking for suggestions about their life on subreddits. In the past few years, many content creators have come forward to create content to help people with their mental health. Many organizations can use stress detection to find which social media users are stressed to help them quickly. So if you want to learn how to use machine learning to detect stress on social media posts, this article is for you. In this article, I will take you through the task of stress detection with machine learning using Python. Stress Detection with Machine Learning Stress detection is a challenging task, as there are so many words that can be used by people on their posts that can show whether a person is having psychological stress or not. While looking for datasets that I can use to train a machine learning model for stress detection, I found a dataset on Kaggle with 116 columns. We only need to use the text and label column for this task. The dataset I am using for this task contains data posted on subreddits related to mental health. This dataset contains various mental health problems shared by people about their life. Fortunately, this dataset is labelled as 0 and 1, where 0 indicates no stress and 1 indicates stress. So in the section below, I will take you through the task of stress detection in social media posts using Python.

Stress Detection using Python Now let's start the task of stress detection with machine learning. I will start this task by importing the necessary Python libraries and the dataset that we need for this task:

```
In [4]: import pandas as pd
import numpy as np
data = pd.read_csv("C:\\Users\\prasann\\Desktop\\DS\\ML Proj\\DataSets_ML Projects\\Stress Detection with Machine Learn
print(data.head())
```

```
subreddit post id sentence range \
0
               ptsd 8601tu
                                  (15, 20)
1
         assistance 81brx9
                                    (0, 5)
2
               ptsd 9ch1zh
                                  (15, 20)
3
      relationships 7rorpp
                                   [5, 10]
  survivorsofabuse 9p2gbc
                                    [0, 5]
                                                text
                                                         id label \
0 He said he had not felt that way before, sugge...
                                                      33181
                                                                 1
1 Hey there r/assistance, Not sure if this is th...
                                                       2606
                                                                 0
2 My mom then hit me with the newspaper and it s...
                                                      38816
                                                                 1
                                                                 1
3 until i met my new boyfriend, he is amazing, h...
                                                        239
4 October is Domestic Violence Awareness Month a...
                                                                 1
                                                       1421
   confidence social timestamp social karma syntax ari ... \
0
          0.8
                     1521614353
                                            5
                                                 1.806818 ...
1
          1.0
                     1527009817
                                                 9.429737 ...
2
          0.8
                                            2
                                                 7.769821
                     1535935605
3
          0.6
                                            0
                                                 2.667798
                     1516429555
4
          0.8
                     1539809005
                                                 7.554238 ...
                                           24
   lex_dal_min_pleasantness lex_dal_min_activation lex_dal_min_imagery \
0
                      1.000
                                                                     1.0
                                             1.1250
1
                      1.125
                                             1.0000
                                                                     1.0
2
                      1.000
                                                                     1.0
                                             1.1429
3
                      1.000
                                             1.1250
                                                                     1.0
4
                      1.000
                                             1.1250
                                                                     1.0
   lex dal avg activation lex dal avg imagery lex dal avg pleasantness \
0
                  1.77000
                                       1.52211
                                                                 1.89556
1
                  1.69586
                                       1.62045
                                                                 1.88919
2
                  1.83088
                                       1.58108
                                                                 1.85828
3
                  1.75356
                                       1.52114
                                                                 1.98848
4
                  1.77644
                                       1.64872
                                                                 1.81456
   social_upvote_ratio social_num_comments syntax_fk_grade sentiment
0
                  0.86
                                          1
                                                              -0.002742
                                                    3.253573
                                          2
1
                  0.65
                                                    8.828316
                                                               0.292857
2
                                          0
                  0.67
                                                    7.841667
                                                               0.011894
3
                  0.50
                                          5
                                                    4.104027
                                                               0.141671
4
                  1.00
                                          1
                                                    7.910952 -0.204167
```

[5 rows x 116 columns]

Let's have a look at whether this dataset contains any null values or not:

So this dataset does not have any null values. Now let's prepare the text column of this dataset to clean the text column with stopwords, links, special symbols and language errors

```
import nltk
In [6]:
        import re
        nltk.download('stopwords')
        stemmer = nltk.SnowballStemmer("english")
        from nltk.corpus import stopwords
        import string
        stopword=set(stopwords.words('english'))
        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 = [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["text"] = data["text"].apply(clean)
```

Now let's have a look at the most used words by the people sharing about their life problems on social media by visualizing a word cloud of the text column:



Stress Detection Model The label column in this dataset contains labels as 0 and 1. 0 means no stress, and 1 means stress. I will use Stress and No stress labels instead of 1 and 0. So let's prepare this column accordingly and select the text and label columns for the process of training a machine learning model:

Now I will split this dataset into training and test sets:

As this task is based on the problem of binary classification, I will be using the Bernoulli Naive Bayes algorithm, which is one of the best algorithms for binary classification problems. So let's train the stress detection model:

Now let's test the performance of our model on some random sentences based on mental health:

So as you can see, we can see go	ood results from our machine	learning model. This is how	v you can train a stress	detection model to
detect stress from social media p	osts. This machine learning r	model can be improved by	feeding it with more da	ita

Summary So this is how you can train a machine learning model to detect stress from social media posts. People often share their feelings on social media platforms. Many organizations can use stress detection to find which social media users are stressed to help them quickly. I hope you liked this article on stress detection with machine learning using Pytho

In []:	
In []:	
In []:	