

Stress Detection with Machine Learning

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Stress is a prevalent issue that can have detrimental effects on an individual's health, productivity, and overall quality of life. However, detecting and managing stress is often challenging because it is a subjective experience and can manifest differently in each person. Traditional methods of stress assessment rely on self-reporting, which can be unreliable and prone to biases.

Machine learning-based stress detection offers several advantages over conventional approaches. It provides an objective and data-driven analysis of stress levels, bypassing the limitations of self-reporting. By automating the stress detection process, individuals can receive timely interventions, leading to better stress management and improved well-being. Now I will walk through the task of stress detection with machine learning 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 [1]: import pandas as pd  
import numpy as np
```

```
In [2]: data = pd.read_csv("stress.csv")
print(data.head())
```

```

      subreddit post_id sentence_range \
0          ptsd 8601tu      (15, 20)
1    assistance 8lbrx9       (0, 5)
2          ptsd 9ch1zh      (15, 20)
3  relationships 7rorpp     [5, 10]
4 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
3 until i met my new boyfriend, he is amazing, h...   239      1
4 October is Domestic Violence Awareness Month a...  1421      1

      confidence  social_timestamp  social_karma  syntax_ari  ... \
0           0.8      1521614353           5      1.806818  ...
1           1.0      1527009817           4      9.429737  ...
2           0.8      1535935605           2      7.769821  ...
3           0.6      1516429555           0      2.667798  ...
4           0.8      1539809005          24      7.554238  ...

      lex_dal_min_pleasantness  lex_dal_min_activation  lex_dal_min_imagery \
0                1.000                1.1250                1.0
1                1.125                1.0000                1.0
2                1.000                1.1429                1.0
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                3.253573  -0.002742
1                0.65                2                8.828316   0.292857
2                0.67                0                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]
```

```
In [3]: # Check whether this dataset contains any null values or not:
print(data.isnull().sum())
```

```
subreddit          0
post_id            0
sentence_range     0
text               0
id                 0
..
lex_dal_avg_pleasantness  0
social_upvote_ratio  0
social_num_comments  0
syntax_fk_grade     0
sentiment           0
Length: 116, dtype: int64
```

So this dataset does not have any null values.

Let's prepare the text column of this dataset to clean the text column with stopwords, links, special symbols and language errors:

```
In [4]: import nltk
import re
nltk.download('stopwords')
stemmer = nltk.SnowballStemmer("english")
from nltk.corpus import stopwords
import string
stopword=set(stopwords.words('english'))
```

```
[nltk_data] Error loading stopwords: <urlopen error [WinError 10060] A
[nltk_data] connection attempt failed because the connected party
[nltk_data] did not properly respond after a period of time, or
[nltk_data] established connection failed because connected host
[nltk_data] has failed to respond>
```

```
In [5]: 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)
```

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:


```
In [8]: from sklearn.feature_extraction.text import CountVectorizer
from sklearn.model_selection import train_test_split

x = np.array(data["text"])
y = np.array(data["label"])

cv = CountVectorizer()
X = cv.fit_transform(x)
xtrain, xtest, ytrain, ytest = train_test_split(X, y,
                                                test_size=0.33,
                                                random_state=42)
```

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:

```
In [9]: from sklearn.naive_bayes import BernoulliNB
model = BernoulliNB()
model.fit(xtrain, ytrain)
```

Out[9]: BernoulliNB()

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

```
In [10]: user = input("Enter a Text: ")
data = cv.transform([user]).toarray()
output = model.predict(data)
print(output)
```

Enter a Text: I am really busy today
['No Stress']

```
In [11]: user = input("Enter a Text: ")
data = cv.transform([user]).toarray()
output = model.predict(data)
print(output)
```

Enter a Text: Sometimes I feel very uncomfortable like I can't do anything
['Stress']

```
In [12]: user = input("Enter a Text: ")
data = cv.transform([user]).toarray()
output = model.predict(data)
print(output)
```

Enter a Text: Sometimes I feel like I have no one who can stay with me
['Stress']

In []:

