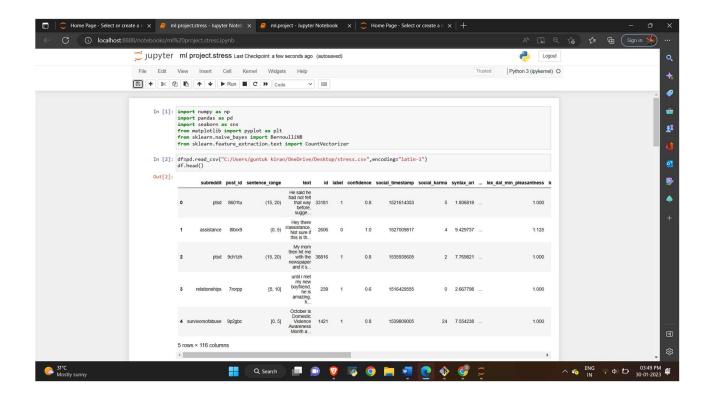
MACHINE LEARNING PROJECT

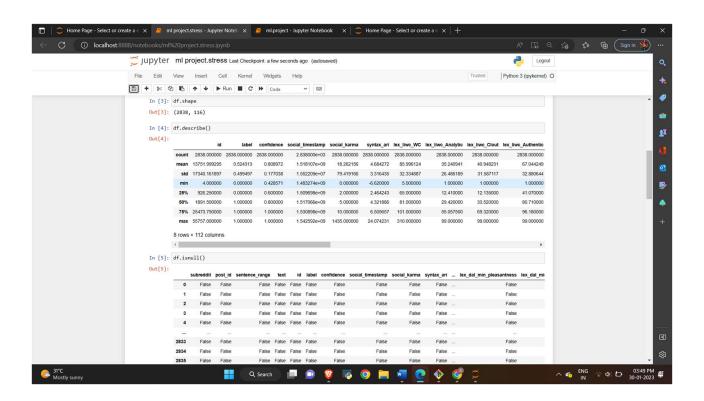
LITERATURE SURVEY

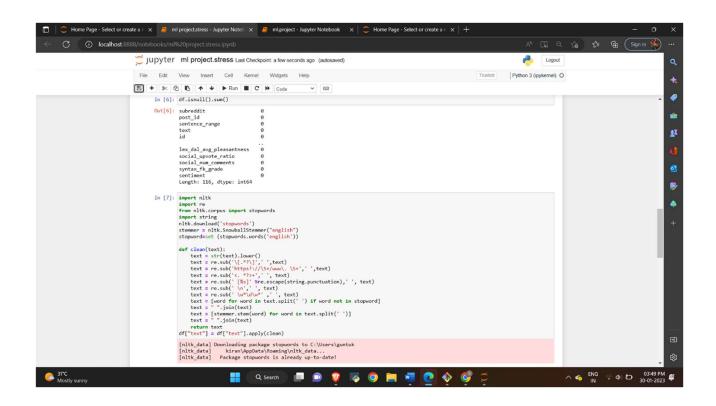
SI.NO.	TITLE	AUTHOR NAME	PUBLISHED YEAR	REMARKS
1	Machine Learning and IoT for Prediction and Detection of Stress	Mr.Purnendu Shekhar Pandey	2017	The paper focuses on using heart rate as a parameter to detect stress in individuals as well as predicting a person's condition to inform them about their health status. To achieve this goal, the authors propose using Machine Learning and the Internet of Things (IoT). IoT is used to communicate a person's health status to them, whereas ML is used to predict their condition. The goal is to help people recognise when they are unhealthy and take action to improve their health before an acute condition occurs.
	Stress Detection with Machine Learning and Deep Learning using Multimodal Physiological Data	Pramod Bobade, Vani M.	2020	In this paper, machine learning and deep learning methods for stress detection are studied. To train the algorithms and identify people who are stressed, the authors employ multimodal physiological data such as heart rate, electrodermal activity, and breathing. The research compares the outcomes produced by each machine learning and deep learning method while discussing their respective efficacies. The authors come to the conclusion that combining multimodal physiological data is useful for stress detection and that deep learning algorithms provide the best results in comparison to traditional machine learning methods.
3	A Decision Tree Optimised SVM Model for Stress Detection using Bio signals	Alana Paul Cruz, Aravind Pradeep, Kavali Riya Sivasankar and Krishnaveni K.S	2020	The paper focuses on the use of decision tree optimization to improve the performance of support vector machine (SVM) models in biosignal stress detection. The authors propose using decision trees to pre-process the biosignal data before training the SVM model with the optimised data. The optimised SVM model's performance is evaluated and compared to traditional SVM models. The proposed decision tree optimised SVM model outperforms traditional SVM models in stress detection using biosignals, according to the results.

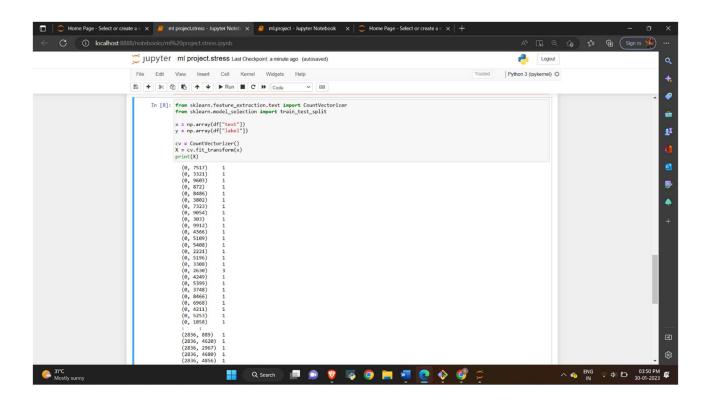
4	Stress detection using deep neural networks	Russell Li and Zhandong Liu		The paper focuses on the use of deep neural networks (DNNs) to detect stress. The authors collect physiological and behavioural data from stress-inducing tasks participants and use this data to train DNNs for stress detection. The DNNs' performance is evaluated and compared to that of traditional machine learning algorithms. DNNs replace traditional machine learning algorithms in stress detection, according to the findings. The authors also examine the DNNs' learned features and discover that they capture complex patterns in physiological and behavioural data that indicate stress. The authors conclude that DNNs are effective at detecting stress and have the potential to provide a deeper understanding of the underlying stress mechanisms.
5	Automatic Stress Detection Using Wearable Sensors and Machine Learning: A Review	Shruti Gedam, Sanchita Paul	2020	In this paper, we make an analysis on automatic stress detection using wearable sensors and machine learning. The authors study the different physiological and behavioural cues, such as heart rate, skin conductance, and movement, that have been employed for stress detection. They also discuss the numerous decision trees, support vector machines, and deep learning machine learning algorithms and methods that have been applied to stress detection. The authors offer recommendations for further research in this field after outlining the possible advantages and disadvantages of utilising wearable sensors and machine learning for stress detection.

Screenshots









```
😊 Home Page - Select or create a 🗈 🗴 🤌 ml project.stress - Jupyter Notelo 🗴 🤌 ml.project - Jupyter Notebook 💢 😂 Home Page - Select or create a 🗈 🗴
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                                             In [9]: xtrain, xtest, ytrain, ytest = train_test_split(X, y, test_size=0.30,random_state=42)
                                           In [12]:
    user=input("Enter the text")
    data=cv.transform([user]).toarray()
    output=model.predict(data)
    print(output)
                                                         Enter the texti am happy
```

Code

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import BernoulliNB\nfrom sklearn.feature extraction.text import
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                      subreddit\n
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38816\n
239\n 1\n 0.6\n 1516429555\n 0\n 2.667798\n ..\n 1.000\n 1.1250\n 1.0\n 1.75356\n 1.52114\n 1.98848\n 0.50\n 5\n 4.104027\n
0.141671\n \n \n 4\n survivorsofabuse\n 9p2qbc\n
[0, 5]\n October is Domestic Violence Awareness Month a...\n 1421\n 1\n 0.8\n 1539809005\n 24\n 7.554238\n
5 \text{ rows} \times 116 \text{ columns}
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relationships 7rorpp [5, 10] \n4 survivorsofabuse 9p2qbc
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33181 1 \n1 Hey there r/assistance, Not sure if this is th...
        0 \n2 My mom then hit me with the newspaper and it s...
1 \n3 until i met my new boyfriend, he is amazing, h...
239 1 \n4 October is Domestic Violence Awareness Month a...
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21.198919\n
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          syntax ari lex liwc WC lex liwc Analytic lex liwc Clout

      4.684272
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False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\n False\
```

False\n	False\n	False\n	False\n	False\n	
False\n	False\n	False\n	False\n	False\n	\n \n
2837\n	False\n	False\n	False\n	False\n	False\n
False\n	False\n	False\n	False\n	False\n	\n
False\n	False\n	False\n	False\n	False\n	
False\n	False\n	False\n	False\n	False\n	\n \n\n

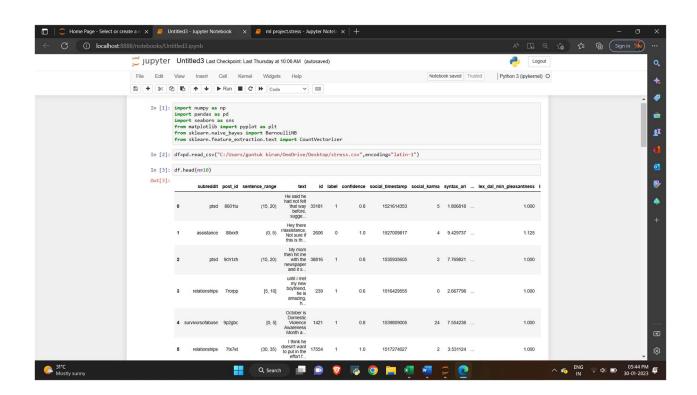
2838 rows × 116 columns

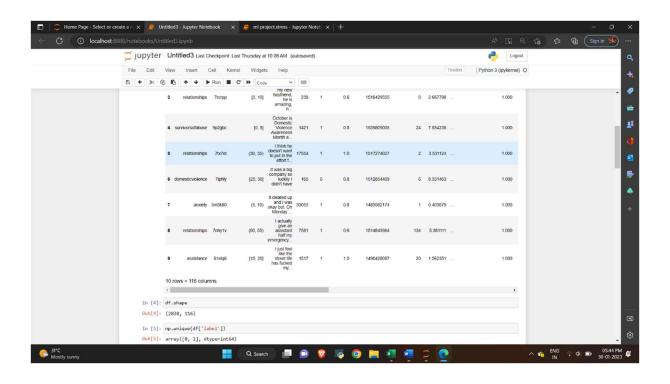
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FalseFalseN1False
False False False \n
False False False \n
False False False \n
\n2833 Falea Falea
False False False False \n2834 False False
False False False False False False
False False False False False False
False False False False \n2837 False False
False False False False False \n2837 False False False False False \n\n social_timestamp social_karma syntax_ari \\\n0 False False False
False False False False False \\\\\\\\\\\\\\\\\\\\\\\\\\\\\\\\\\\\
False \nl False False False
\n2 False False \n3
False False \n4 False
False False \n
\n2833 False False False \n2835
False False \n2836 False False False
False False \n2837 False False False \n\n lex_dal_min_pleasantness lex_dal_min_activation
lex_dal_min_imagery \\\n0 False
False False \n1 False
False False \n2 False
False False \n3 False
False False \n4 False
False False \n
\n2833 False
False False \n2834 False False
False \n2836 False
False \n2837 False
False
<pre>lex_dal_avg_imagery lex_dal_avg_pleasantness \\\n0</pre>
False False \n1
False False \n2
False False \n3
False False \n4
False False \n
\n2833
False False Talse \n2834
False False \n2835
False False h2836
False False False \n2837

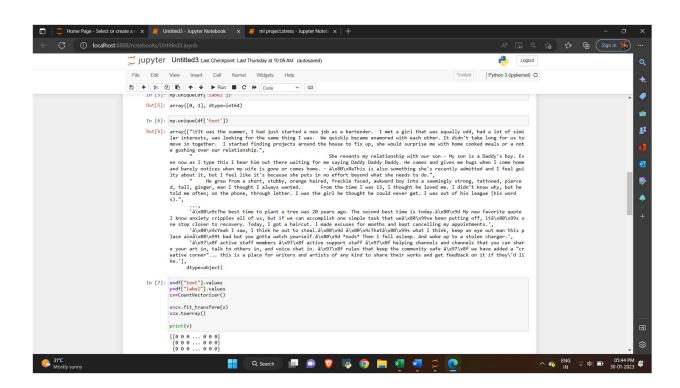
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False
                                                    False
                                                                                                                      False \n\n
social upvote ratio social num comments syntax fk grade sentiment
\n0
                                                      False
                                                                                                           False
                                                                                                                                                       False
False \n1
                                                                         False
                                                                                                                             False
False False \n2
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False
                        False \n3
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                        False \n...
False
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                                                                                                                                                              . . .
                          ...\n2833
                                                                                               False
                                                                                                                                                    False
False
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                                                                                                    False
                                                                                                                                                         False
False
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False
                                                                                                    False
                                                                                                                                                         False
False
                        False \n2837
                                                                                                    False
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                                                                              0\nsocial num comments
0\nsyntax fk grade
                                                                              0\nsentiment
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nltk.SnowballStemmer(\"english\") \nstopword=set
(stopwords.words('english')) \n\ndef clean(text):\n
str(text).lower() \ text = re.sub('\\[.*?\\]',' ',text) \ n
', text) \n text = re.sub(' [%s]' %re.escape(string.punctuation),' ',
text) \n text = re.sub(' \\n',' ', text) \n text = re.sub('
\w^*\d^w^' ,' ', text)\n text = [word for word in text.split(' ')
if word not in stopword]\n text = \" \".join(text)\n text =
[stemmer.stem(word) for word in text.split(' ')]\n text = \"
\".join(text)\n return text\ndf[\"text\"] =
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kiran\\AppData\\Roaming\\nltk data...\n[nltk data] Package stopwords
is already up-to-
date!\n"}]}, {"metadata": {"trusted": false}, "id": "d9645164", "cell type":"
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cv.fit transform(x)\nprint(X)","execution count":8,"outputs":[{"name":"
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(0, 9603) \times 1^{n} (0, 872) \times 1^{n} (0, 8486) \times 1^{n} (0, 3802) \times 1^{n} (0, 9603) \times
7323)\t1\n (0, 9054)\t1\n (0, 303)\t1\n (0, 9912)\t1\n (0,
4366)\t1\n (0, 5109)\t1\n (0, 5408)\t1\n (0, 2221)\t1\n (0,
5196) \t1\n (0, 3308) \t1\n (0, 2630) \t3\n (0, 4249) \t1\n (0, 60, 60) \t3\n
5399) \t1\n (0, 3748) \t1\n (0, 8466) \t1\n (0, 6968) \t1\n 
4211)\t1\n (0, 5253)\t1\n (0, 1858)\t1\n :\t:\n (2836, 889)\t1\n
(2836, 4620) \t1\n (2836, 2967) \t1\n (2836, 4680) \t1\n (2836, 4680)
4856)\t1\n (2836, 4576)\t1\n (2837, 7517)\t2\n (2837, 3057)\t1\n
```

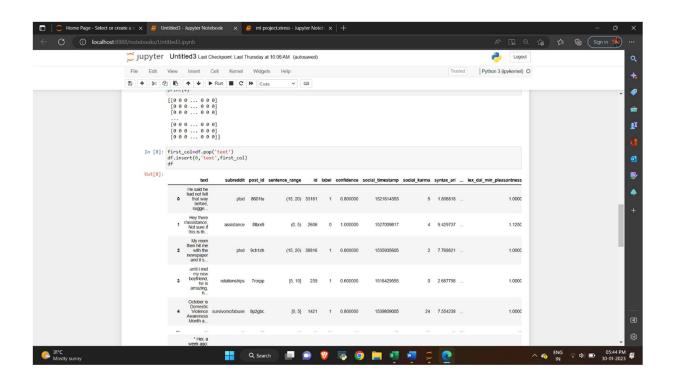
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6876)\t1\n (2837, 4381)\t1\n (2837, 9828)\t1\n (2837, 5657)\t1\n
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)\nprint(output)","execution count":12,"outputs":[{"name":"stdout","out
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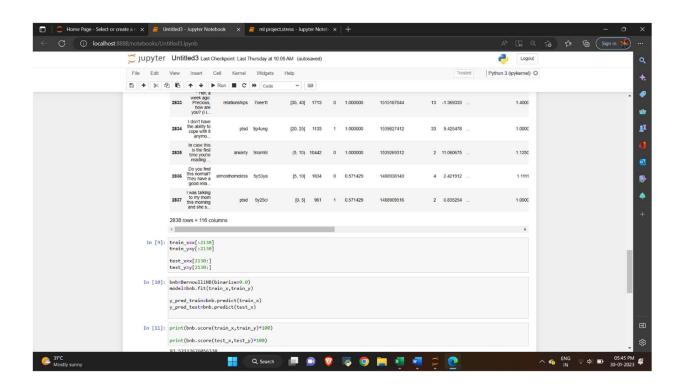
Sample code on stress file

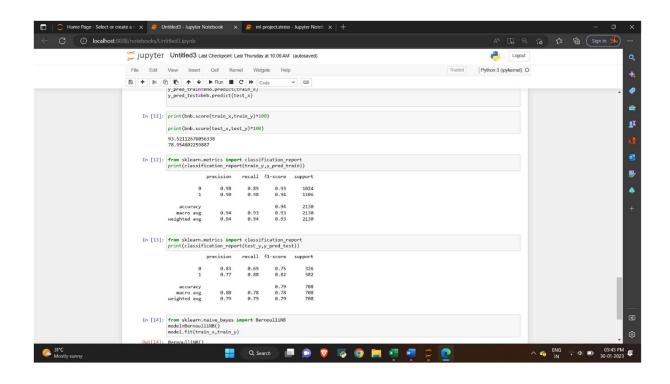


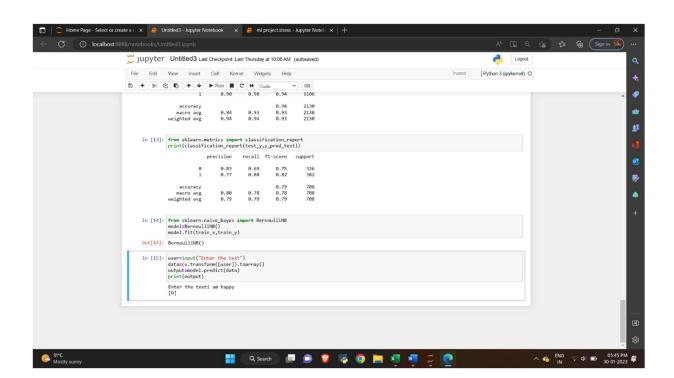












```
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import BernoulliNB\nfrom sklearn.feature extraction.text import
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ptsd 8601tu
                  (15, 20)
                           \n1
                                      assistance 8lbrx9
5)
    \n2
                     ptsd 9ch1zh
                                       (15, 20)
                                                   \n3
relationships 7rorpp [5, 10] \n4 survivorsofabuse 9p2gbc [0, 5] \n5 relationships 7tx7et (30, 35) \n6
relationships 7rorpp
                              [25, 30]
domesticviolence 7iphly
                                         \n7
                                                       anxiety 5m3k80
(5, 10) \n8
                 relationships 7nhy1v
                                             (50, 55)
                                                      \n9
assistance 61eiq6
                        [15, 20] \ \n\n
       id label \\n0 He said he had not felt that way before,
sugge... 33181
                   1 \n1 Hey there r/assistance, Not sure if this
                       \n2 My mom then hit me with the newspaper and
         2606
                    0
is th...
it s... 38816
                   1 \n3 until i met my new boyfriend, he is
                239
                      1 \n4 October is Domestic Violence
amazing, h...
                             1 \n5 I think he doesn't want to
Awareness Month a... 1421
put in the effort f... 17554
                                 1 \n6 It was a big company so
luckily I didn't have ... 165
                                     0
                                         \n7 It cleared up and I
was okay but. On Monday ... 33053
                                       1
                                         \n8 I actually give an
assistant half my emergency... 7581
                                          1 \n9 I just feel like
the street life has fucked my... 1517
                                            1
                                               \n\n
                                                       confidence
                                          ... \\\n0
social timestamp social karma syntax ari
                                                              0.8
                     5 1.806818 ...
1521614353
                                           \n1
                                                       1.0
                                                       0.8
1527009817
                           9.429737 ...
                                           n2
                           7.769821 ...
1535935605
                     2
                                           \n3
                                                       0.6
1516429555
                     0
                           2.667798
                                           \n4
                                                       0.8
                                     . . .
                     24
                                           \n5
1539809005
                           7.554238
                                                       1.0
                                     . . .
1517274027
                     2
                           3.531124
                                           \n6
                                                       0.8
                                     . . .
                           8.331463 ...
1512854409
                      6
                                           \n7
                                                       0.8
                     1
1483582174
                           0.403679 ...
                                           \n8
                                                       0.6
                           5.381111 ...
1514843984
                    134
                                           \n9
                                                       1.0
1490428087
                     20
                           1.562351 ...
                                           \n\n
lex dal min pleasantness lex dal min activation lex dal min imagery
\\\n0
                         1.000
                                                1.1250
1.0
     \n1
                             1.125
                                                    1.0000
1.0
     n2
                             1.000
                                                    1.1429
1.0
     \n3
                             1.000
                                                    1.1250
1.0
     \n4
                             1.000
                                                    1.1250
1.0
     \n5
                             1.000
                                                    1.1818
1.0
                             1.000
     \n6
                                                    1.1250
1.0
     \n7
                             1.000
                                                    1.1429
1.0
                             1.000
                                                    1.0000
     \n8
1.0
     \n9
                             1.000
                                                    1.1250
1.0
     \n\n lex dal avg activation lex dal avg imagery
lex_dal_avg_pleasantness \\\n0
                                               1.77000
1.52211
                         1.89556
                                   \n1
                                                       1.69586
1.62045
                         1.88919
                                   \n2
                                                       1.83088
```

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1.58108
                                                                     1.85828 \n3
                                                                                                                                                            1.75356
                                                                      1.98848 \n4
                                                                                                                                                          1.77644
 1.52114
                                                                     1.81456 \n5
1.82046 \n6
                                                                                                                                                           1.70415
 1.64872
 1.44396
                                                                                                                                                            1.66985
 1.57757
                                                                     1.89926 \n7
                                                                                                                                                            1.69549
 1.44860
                                                                     1.88963 \n8
                                                                                                                                                            1.73571
                                       1.93607 \n9 1.74950
1.89320 \n\n social_upvote_ratio
 1.71471
 1.67353
 social num comments syntax fk grade sentiment \n0

      social_num_comments
      syntax_fk_grade
      sentiment
      \n0

      0.86
      1
      3.253573
      -0.002742
      \n1

      0.65
      2
      8.828316
      0.292857
      \n2

      0.67
      0
      7.841667
      0.011894
      \n3

      0.50
      5
      4.104027
      0.141671
      \n4

      1.00
      1
      7.910952
      -0.204167
      \n5

      0.75
      5
      4.338429
      -0.550000
      \n6

      0.88
      29
      8.394244
      0.277381
      \n7

      1.00
      0
      1.028654
      -0.079630
      \n8

      0.94
      43
      5.991049
      -0.272917
      \n9

      0.80
      6
      2.851935
      -0.117262
      \n\r

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sentence_range\n text\n id\n label\n confidence\n

social_timestamp\n social_karma\n syntax_ari\n ...\n
 lex_dal_min_pleasantness\n lex_dal_min_activation\n
lex_dal_min_imagery\n lex_dal_avg_activation\n lex_dal_avg_imagery\n lex_dal_avg_pleasantness\n social_upvote_ratio\n social_num_comments\n syntax_fk_grade\n
(0, 5)\n Hey there r/assistance, Not sure if this is th...\n
20)\n My mom then hit me with the newspaper and it s...\n
 38816\n 1\n 0.8\n 1535935605\n 2\n 7.769821\n ...\n 1.000\n 1.1429\n 1.0\n 1.83088\n
..\n 1.000\n 1.1250\n 1.0\n 1.75356\n 1.52114\n 1.98848\n 0.50\n 5\n 4.104027\n 0.141671\n \n \n 4\n survivorsofabuse\n 9p2gbc\n
[0, 5]\n October is Domestic Violence Awareness Month a...\n 1421\n 1\n 0.8\n 1539809005\n 24\n 7.554238\n
1421\lambda 1\lambda 0.8\lambda 1339809003\lambda 24\lambda 7.334236\lambda \lambda \lambda 1.0\lambda 1.77644\lambda 1.64872\lambda 1.81456\lambda 1.00\lambda 1\lambda 7.910952\lambda -0.204167\lambda \lambda \lam
 17554\n 1\n 1.0\n 1517274027\n 2\n 3.531124\n
 ...\n 1.000\n 1.1818\n 1.0\n 1.70415\n 1.44396\n 1.82046\n 0.75\n 5\n 4.338429\n
```

```
0.550000\n \n \n 6\n domesticviolence\n 7iphly\n [25, 30]\n It was a big company so luckily I didn't have ...\n
       0\n 0.8\n 1512854409\n 6\n 8.331463\n
165\n
         1.000\n 1.1250\n 1.0\n
                                        1.66985\n
...\n
1.57757\n
           1.89926\n 0.88\n 29\n 8.394244\n
          \n \n
                      7\n anxiety\n
0.277381\n
                                          5m3k80\n
      It cleared up and I was okay but. On Monday ...\n
33053\n
        1\n 0.8\n 1483582174\n 1\n 0.403679\n
       1.000\n
                 1.1429\n 1.0\n
...\n
                                       1.69549\n
1.44860\n 1.88963\n 1.00\n 0\n
                                        1.028654\n
0.079630\n
0.079630\n \n \n 8\n relationships\n 7nhy1v\n (50, 55)\n I actually give an assistant half my emergency...\n
7581\n
        1\n 0.6\n 1514843984\n 134\n 5.381111\n
...\n
        1.000\n 1.0000\n 1.0\n
                                       1.73571\n
         1.93607\n 0.94\n 43\n
1.71471\n
                                          5.991049\n
           \n \n 9\n assistance\n 61eiq6\n
0.272917\n
[15, 20]\n I just feel like the street life has fucked my...\n
1517\n
        1\n 1.0\n 1490428087\n 20\n
...\n
         1.000\n 1.1250\n 1.0\n
                                        1.74956\n
1.67353\n
         1.89320\n 0.80\n 6\n
                                       2.851935\n
0.117262\n
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10 rows × 116 columns

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a new job as a bartender. I met a girl that was equally odd, had a lot
of similar interests, was looking for the same thing I was. We quickly
became enamored with each other. It didn't take long for us to move in
together. I started finding projects around the house to fix up, she
would surprise me with home cooked meals or a note qushing over our
relationship.\",\n \"
She resents my relationship with our son - My son is a Daddy's boy.
Even now as I type this I hear him out there waiting for me saying
Daddy Daddy Daddy. He comes and gives me hugs when I come home and
barely notices when my wife is gone or comes home. - \hat{a}\x80\\x8eThis is
also something she's recently admitted and I feel quilty about it, but
I feel like it's because she puts in no effort beyond what she needs to
do.\",\n \" He grew from a short, stubby, orange haired,
freckle faced, awkward boy into a seemingly strong, tattooed, pierced,
tall, ginger, man I thought I always wanted. From the time I was
13, I thought he loved me. I didn't know why, but he told me often; on
the phone, through letter. I was the girl he thought he could never
get. I was out of his league (his words).\",\n
'â\\x80\\x9cThe best time to plant a tree was 20 years ago. The second
```

```
best time is today.â\\x80\\x9d My new favorite quote I know anxiety
cripples all of us, but if we can accomplish one simple task that
\text{weâ}\x80\x99\text{ve} been putting off, itâ\x80\x99s one stop closer to
recovery. Today, I got a haircut. I made excuses for months and kept
cancelling my appointments.',\n 'a\\x80\\x9cYeah I saw, I think
he out to steal.\hat{a}\x80\\x9d \hat{a}\x80\\x9cThat\hat{a}\x80\\x99s what I think,
keep an eye out man this place ainâ\\x80\\x99t bad but you gotta watch
yourself.\hat{a}\x80\\x9d *nods* Then I fell asleep. And woke up to a stolen
                \hat{a}\x97\x8f active staff members \hat{a}\x97\x8f active
charger.',\n
support staff \hat{a} \times 97 \times 8f helping channels and channels that you can
share your art in, talk to others in, and voice chat in. \hat{a} \times 97 \times 6
rules that keep the community safe \hat{a} \times 97 \times 8f we have added a
\"creative corner\"... this is a place for writers and artists of any
kind to share their works and get feedback on it if they \\'d like.'], \n
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text
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                                ptsd \n1 Hey there r/assistance,
before, sugge...
Not sure if this is th...
                                  assistance
                                                \n2
                                                          My mom then hit
me with the newspaper and it s...
                                                  ptsd
                                                           \n3
                                                                   until i
                                                                  \n4
met my new boyfriend, he is amazing, h... relationships
October is Domestic Violence Awareness Month a... survivorsofabuse
... \n2833 * Her, a week ago: Precious, how are you? (I i...
relationships \n2834 I don't have the ability to cope with it
                      ptsd \n2835 In case this is the first time
anymo...
                              anxiety \n2836 Do you find this normal?
you're reading ...
They have a good rela... almosthomeless \n2837 I was talking to my mom this morning and she s... ptsd \n\n post_id
sentence range id label confidence social timestamp \\\n0
8601tu (15, 20) 33181 1 0.800000 1521614353
                                                                          \n1

      (0, 5)
      2606
      0
      1.000000

      (15, 20)
      38816
      1
      0.800000

      [5, 10]
      239
      1
      0.600000

      [0, 5]
      1421
      1
      0.800000

8lbrx9
                                                           1527009817
                                                                          \n2
9ch1zh
                                                          1535935605
                                                                          \n3
7rorpp
                                                          1516429555
                                                        1539809005

      9p2gbc
      [0, 5]
      1421
      1
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      1539809

      \n...
      \n2833
      70eelt
      [35, 40]
      1713
      0
      1.000000

9p2gbc
1515187044 \n2834 9p4ung [20, 25] 1133 1 1.000000
                                    (5, 10) 10442
1539827412 \n2835 9nam61
                                                           0
                                                                1.000000
             \n2836 5y53ya [5, 10] 1834
\n2837 5y25cl [0, 5] 961
1539269312
                                                          0
                                                                0.571429
1488938143
                                                          1
                                                                0.571429
1488909516 \n\n social_karma syntax_ari ...
lex dal min pleasantness \\\n0
                                               5 1.806818 ...
4
                                    9.429737 ...
1.1250 \n2
                              2
                                   7.769821 ...
1.0000 \n3
                             0
                                    2.667798 ...
1.0000
         \n4
                           24
                                    7.554238 ...
1.0000 \n...
                13 -1.369333 ...
... \n2833
                                                                     1.4000
```

```
\n2834 33 9.425478 ...
\n2835 2 11.060675 ...
\n2836 4 2.421912 ...
\n2837 2 0.835254 ...
                                                        1.0000
                                                        1.1250
                                                        1.1111
                                                        1.0000 \n\n
lex_dal_min_activation lex_dal_min_imagery lex_dal_avg_activation
       1.1250 1.0000 1.0
\\\n0
1.77000 \n1
                                 1.1429
1.69586 \n2
                                                         1.0
1.83088 \n3
                                 1.1250
                                                         1.0
1.75356 \n4
1.77644 \n...
\n2833
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1.79768 \n2836
1.71642 \n2837
                                 1.1250
                                                         1.0
                                 1.1429
                                                         1.0
                                1.0000
                                                         1.0
1.71642 \n2037

1.68891 \n\n lex_dal_avg_imagery lex_dal_avg_pleasantness
social_upvote_ratio \\\n0 \\\n0 \\.1.52211
1.89556
                       0.86 \ \n1
                                                 1.62045
                                          ...
\n \n \n \t ext\n \subreddit\n \post id\n
sentence_range\n id\n label\n confidence\n social_timestamp\n social_karma\n syntax_ari\n ...\n
lex dal min pleasantness\n lex_dal_min_activation\n
lex_dal_min_preasantness\n lex_dal_min_activation\n lex_dal_min_imagery\n lex_dal_avg_activation\n lex_dal_avg_imagery\n lex_dal_avg_pleasantness\n social_upvote_ratio\n social_num_comments\n syntax_fk_grade\n sentiment\n \n \n \n \n \n \n \n \n \tag{601tu\n} (15, 20)\n 33181\n 1\n 0.800000\n 1521614353\n 5\n 1.806818\n ...\n 1.0000\n 1.1250\n 1.0\n 1.7000\n 1.52211\n 1.89556\n 0.66\n 1.0\n
```

```
My mom then hit me with the newspaper and it s...\n ptsd\n
9ch1zh\n (15, 20)\n 38816\n 1\n 0.800000\n 1535935605\n 2\n 7.769821\n ...\n 1.0000\n
1535935605\n 2\n 7.769821\n ...\n 1.1429\n 1.0\n 1.83088\n 1.58108\n
                                                            1.0000\n
1.85828\n
0.67\n
             0\n 7.841667\n 0.011894\n \n \n 3\n
until i met my new boyfriend, he is amazing, h...\n
relationships\n 7rorpp\n [5, 10]\n 239\n 1\n 0.600000\n 1516429555\n 0\n 2.667798\n ...\n
1.0000\n
1.98848\n
               1.1250\n 1.0\n 1.75356\n 1.52114\n
      48\n 0.50\n 5\n 4.104027\n 0.141671\n \n 4\n October is Domestic Violence Awareness Month a...\n
survivorsofabuse\n 9p2gbc\n [0, 5]\n 1421\n 1\n
0.800000\n 1539809005\n 24\n 7.554238\n ...\n
-0.204167\n \n
Precious, how are you? (I i...\n relationships\n 7oee1t\n
[35, 40]\n 1713\n 0\n 1.000000\n 1515187044\n 13\n -1.369333\n ...\n 1.4000\n 1.00000\n 1.0000\n 1.0\n 1.71133\n 1.45301\n 2.00304\n 0.84\n 16\n 0.254444\n 0.552066\n \n \n \n 2834\n I don't have the ability to cope with it anymo...\n ptsd\n 9p4ung\n [20, 25]\n 1133\n 1\n 1.00000\n 1.0000\n 1.0\n 33\n 9.425478\n ...\n 1.0000\n 1.0000\n 1.0\n
1.65003\n 1.56842\n 1.81527\n 0.96\n 6\n
8.640664\n -0.220370\n \n \n 2835\n In case this is the first time you're reading ...\n anxiety\n 9nam61\n
                                                                  In case this
(5, 10)\n 10442\n 0\n 1.000000\n 1539269312\n 2\n 11.060675\n ...\n 1.1250\n 1
2\n 11.060675\n
Do you find
5y53ya\n [5, 10]\n 1834\n 0\n 0.571429\n 1488938143\n 4\n 2.421912\n ...\n 1.1111\n 1.1429\n 1.0\n 1.71642\n 1.57627\n 1.89972\n 0.75\n 7\n 4.036765\n 0.159722\n \n \n 2837\n
I was talking to my mom this morning and she s...\n ptsd\n
5y25c1\n [0, 5]\n 961\n 1\n 0.571429\n
1488909516\n 2\n 0.835254\n ...\n 1.0000\n 1.0000\n 1.0000\n 1.68891\n 1.44615\n 1.89707\n
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2838 rows × 116 columns

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1106\n\n
                                                 0.94
                                                          2130\n
          accuracy
                                                                    macro
avq
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                                                         0.88
                                                                    0.82
382\n\n
          accuracy
                                                0.79
                                                           708\n
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