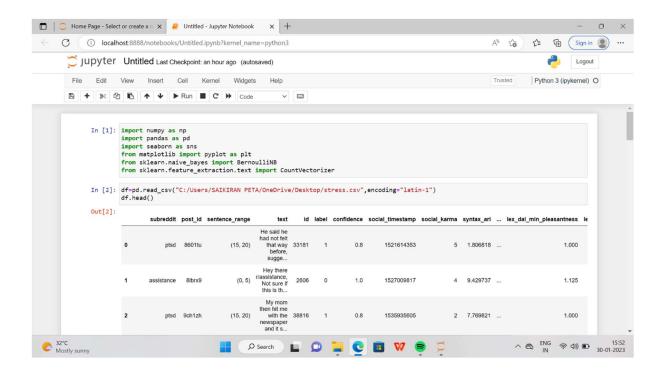
MACHINE LEARNING PROJECT

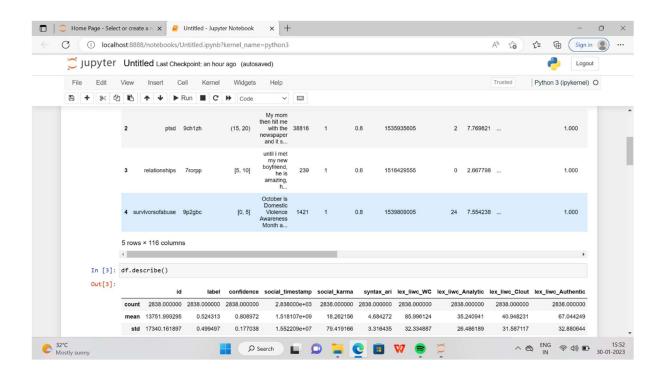
LITERATURE SURVEY

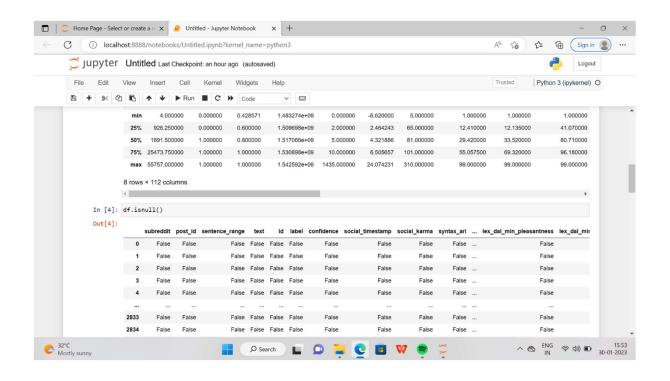
SI.NO.	TITLE	AUTHOR NAME	PUBLISHED YEAR	REMARKS
1	Machine Learning and IoT for Prediction and Detection of Stress	Mr.Purnendu Shekhar Pandey	2017	The study focuses on using heart rate as a measure to identify stress in people and to forecast a person's state in order to let them know how their health is doing. Internet of Things (IoT) and machine learning are the methods the authors suggest adopting to achieve this. While ML is used to foretell a person's condition, IoT is utilised to inform that individual of their health status. In order to prevent an acute disease from developing, it is important for people to learn how to detect when they are in an unhealthy state and take steps to get better.
2	Automatic Stress Detection Using Wearable Sensors and Machine Learning: A Review	Shruti Gedam, Sanchita Paul	2020	The paper provides a review of existing research on the use of wearable sensors and machine learning for automatic stress detection. The authors survey the different types of physiological and behavioural signals that have been used for stress detection, including heart rate, skin conductance, and movement. They also go over the different machine learning algorithms and techniques used for stress detection, such as decision trees, support vector machines, and deep learning. The authors conclude by discussing the potential benefits and limitations of using wearable sensors and machine learning for stress detection, as well as future research directions.
3	Stress Detection with Machine Learning and Deep Learning using Multimodal Physiological Data	Vani M.	2020	The paper discusses the use of machine learning and deep learning algorithms in the detection of stress. The authors train the algorithms and detect stress in individuals using multimodal physiological data such as heart rate, electrodermal activity, and respiration. The paper discusses the efficacy of various machine learning and deep learning methods and compares the results obtained by each. The authors conclude that using multimodal physiological data for stress detection is effective, and that deep learning algorithms easily surpass traditional machine learning methods.

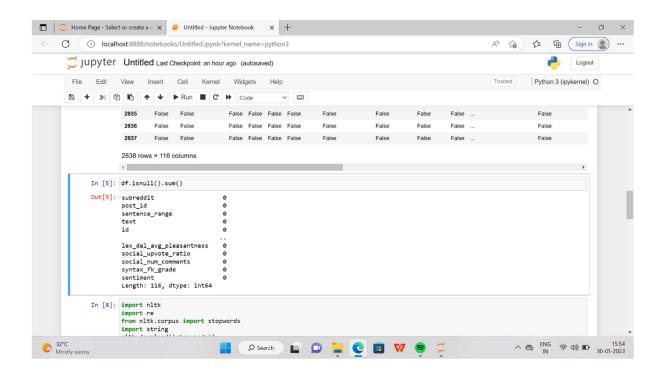
4	A Decision Tree Optimised SVM Model for Stress Detection using Biosignals	Alana Paul Cruz, Aravind Pradeep, Kavali Riya Sivasankar and Krishnaveni K.S	2020	The main topic of the research is the application of decision tree optimization to enhance the performance of support vector machine (SVM) models in biosignal-based stress detection. The authors suggest pre-processing the biosignal data with decision trees and then training the SVM model with the refined data. In comparison to standard SVM models, the performance of the optimised SVM model is assessed. The results show that in stress detection utilising biosignals, the proposed decision tree optimised SVM model outperforms traditional SVM models.
5	Stress detection using deep neural networks	Russell Li and Zhandong Liu		The paper focuses on detecting stress using deep neural networks (DNNs). In order to train DNNs for stress detection, the scientists collect physiological and behavioural data from volunteers who complete stressful tasks. The effectiveness of the DNNs is assessed and contrasted with that of more established machine learning techniques. The findings demonstrate that DNNs perform better in stress detection than conventional machine learning algorithms. The authors also examine the DNNs' learnt features and discover that they can recognise intricate patterns in physiological and behavioural data that point to stress. The authors come to the conclusion that DNNs are useful for spotting stress and may help us understand its underlying causes better.

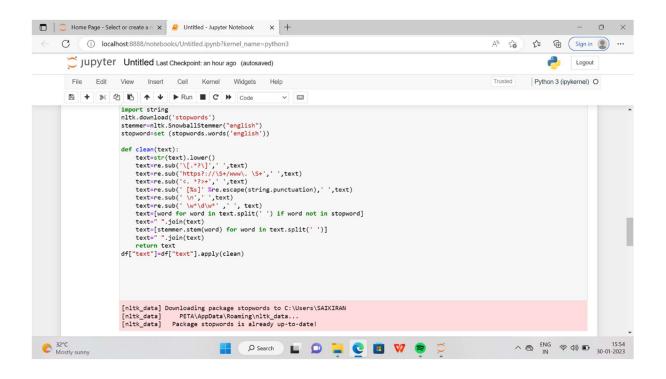
SCREENSHOTS

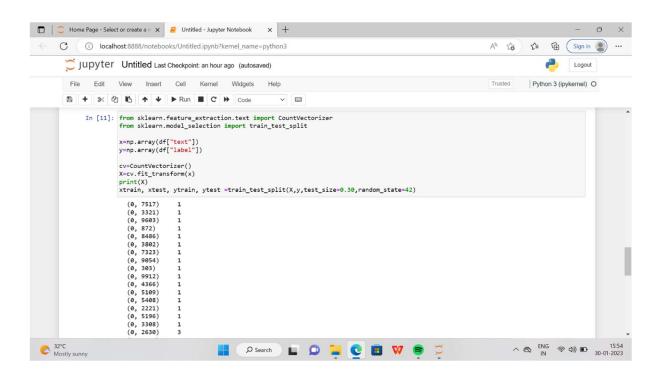


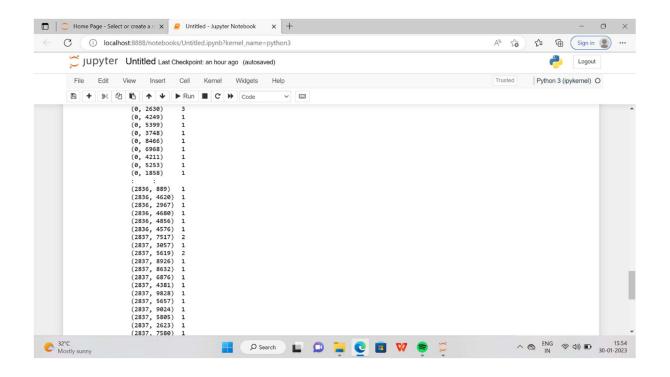


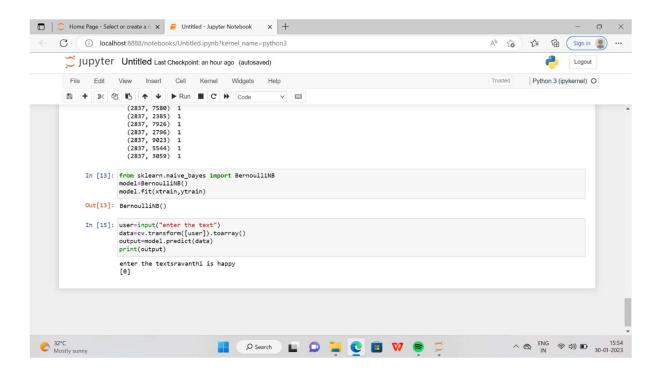












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                                      [5, 10]
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20) \n3
           relationships 7rorpp
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9p2qbc
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text
       1 \n1 Hey there r/assistance, Not sure if this is th... 2606
  \n2 My mom then hit me with the newspaper and it s... 38816
                                                           1 \n4
\n3 until i met my new boyfriend, he is amazing, h... 239
October is Domestic Violence Awareness Month a... 1421 1 \n\
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2 7.769821 ...
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social num comments\n syntax fk grade\n sentiment\n \n \n \n
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th...\n 2606\n 0\n 1.0\n 1527009817\n 4\n 9.429737\n ...\n 1.125\n 1.0000\n 1.0\n 1.
0.292857\n
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hit me with the newspaper and it s...\n 38816\n 1\n 1535935605\n 2\n 7.769821\n ...\n 1.000\n 1.000\n
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7rorpp\n [5, 10]\n until i met my new boyfriend, he is amazing,
h...\n 239\n 1\n 0.6\n 1516429555\n 0\n
         ...\n 1.000\n 1.1250\n 1.0\n 1.75356\n 1.98848\n 0.50\n 5\n 4.104027\n 0.141673
2.667798\n
1.52114\n
                                                     0.141671\n
           4\n survivorsofabuse\n 9p2gbc\n [0, 5]\n
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      1.721430
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      1.751760
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lex dal avg imagery lex dal avg pleasantness social upvote ratio \\\ncount
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False ... \n2833 False False False False False False ... \n2835 False False ... \n2836 False False ... \n2837 False False False False False False ... \n/n
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text=str(text).lower()\n
text=re.sub('https?://\\S+/www\\. \\S+',' ',text)\n
                                                       text=re.sub('+','
',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
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                 text=[stemmer.stem(word) for word in text.split(' ')]\n
text=\" \".join(text)\n
                           return
text\ndf[\"text\"]=df[\"text\"].apply(clean)\n
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THE END