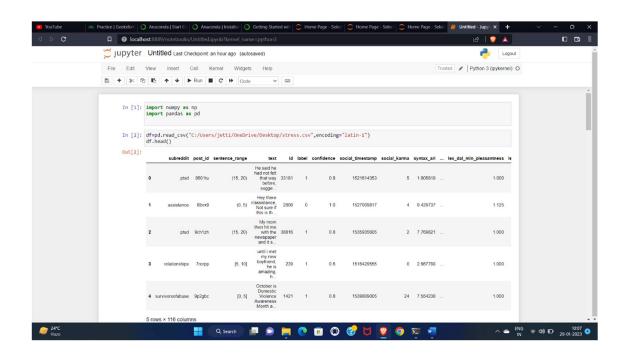
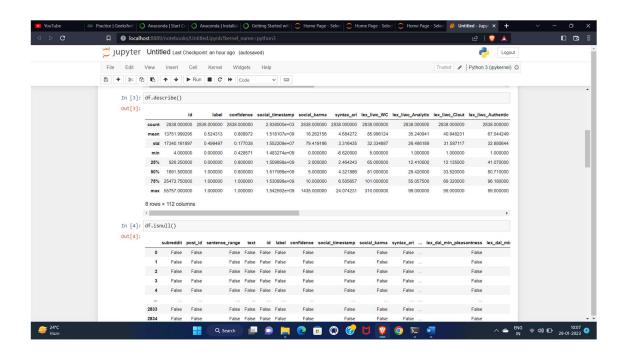
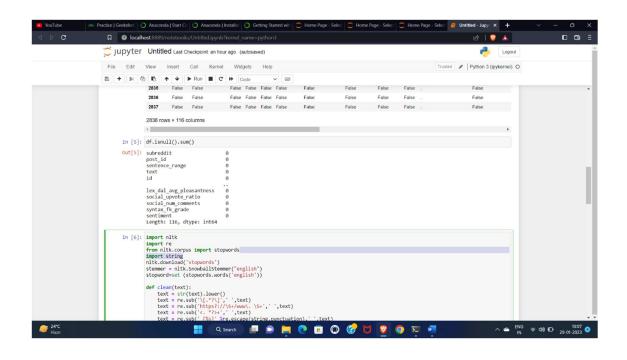
Sr. No	Title of paper	Name of Authors	Published year	Remarks
1.	Stress detection using deep neural networks	Russell Li and Zhandong Liu	2020	The need for manually created features is a drawback of conventional machine learning techniques. If features are incorrectly identified, accuracy declines. To address this deficiency, we developed two deep neural networks 1-dimensional (1D) convolutional neural network (chest worn sensor), it achieved 99.80% and 99.55% accuracy and multilayer perceptron neural network(wrist worn sensor), it achieved
2.	Machine Learning and IoT for Prediction and Detection of Stress.	Mr.Purnendu Shekhar Pandey BML	2017	99.65% and 98.38% accuracy. Algorithms: Logistic Regression. Support Vector Machine (SVM). VF - 15 algorithm. Naive Bayes. VF - 15 with weights to features. The created prototype uses a person's heart rate variability to determine whether they are under stress.
3.	Stress Detection with Machine Learning and Deep Learning using Multimodal Physiological Data	Pramod Bobade and Vani M.	2020	Algorithms like K- Nearest Neighbour, Linear Discriminant Analysis, Random Forest, Decision Tree, AdaBoost and Kernel Support Vector Machine are used to compare the binary classifications and accuracies for three- class. The used dataset is WESAD dataset contains data from

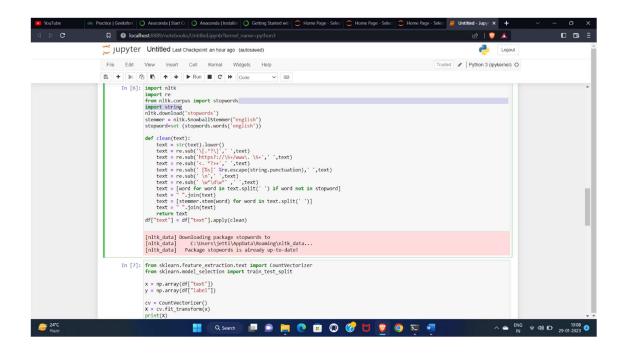
4.	A Decision Tree Optimised SVM Model for Stress Detection using Biosignals	Alana Paul Cruz, Aravind Pradeep, Kavali Riya Sivasankar and Krishnaveni K.S	2020	multiple physiological modalities like three-axis acceleration (ACC), respiration (RESP), electrodermal activity (EDA), electrocardiogram (ECG), body temperature (TEMP), electromyogram (EMG) and blood volume pulse (BVP) which is not available in other datasets, which makes this work suitable for the detection of stress in human being. This model has achieved the accuracy of 84.32% and 95.21% on a three-class and a binary classification problems Test study was directed and substantiated for stress detection using database "drivedb" [Stress Recognition in Automobile Drivers] which was taken from the website Physionet. Initially, the model was trained using Cubic SVM with Gaussian Kernel. For a better model, here we have used Tree Optimised SVM which is a combination of Decision Tree and SVM algorithms. The classification is done using various QRS detection algorithms and other functions in MATLAB.
5.	Automatic Stress Detection Using Wearable Sensors and Machine Learning	Shruti Gedam and Sanchita Paul	2020	In this various methods of stress detection is explained.

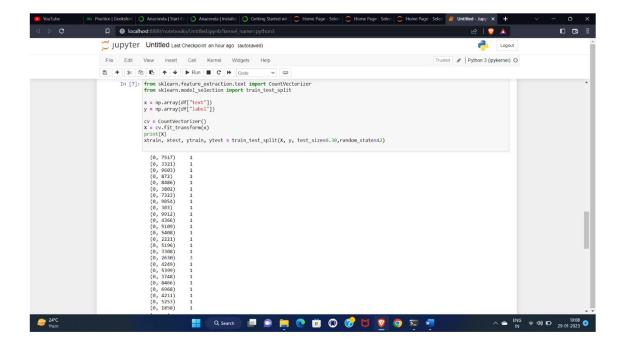
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IOT devices.	u
2.Stress Detection	
through Physiologica	al
Signals	
3.Stress Detection Us	sing
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4.Stress Detection Us	sing
Videos	
5.Stress Detection in	1
Various Environment	ts
using Wearable Sens	ors
Support vector	
machine, Random	
forest and K-Nearest	:
Neighbour are the m	nost
effective classificatio	
algorithms.	
The drawback of	this
study is the extensive	
usage of several	
features that were	
associated with one	
another by the	
researchers, which	
lengthened	
computation times.	
Additionally, some of	f
them collected	1
physiological signals	
using expensive	
commercial equipme	
when it was possible	e to
use	
inexpensive sensors.	•

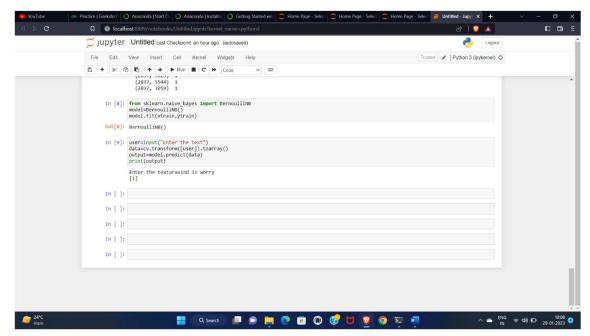












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text id label \\n0 He said he had not felt that way before, sugge... 33181 1 \n1 Hey there r/assistance, Not sure if this is th... 2606 0 \n2 My mom then hit me with the newspaper and it s... 38816 1 \n3 until i met my new boyfriend, he is amazing, h...
239 1 \n4 October is Domestic Violence Awareness Month a... 1421
1 \n\n confidence social_timestamp social_karma syntax_ari ...
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      1.77644
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                                         1.69586
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{\tt 1.81456} \qquad \verb|\n\n| \qquad {\tt social\_upvote\_ratio} \qquad {\tt social\_num\_comments} \qquad {\tt syntax\_fk\_grade}

    sentiment
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    1
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    -0.002742
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    0.65
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                                                                             8.82831

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My mom then hit me with the newspaper and it s...\n 38816\n 1\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 1.58108\n 1.85828\n 0.67\n 0\n 7.841667\n 0.011894\n \n \n 3\n relationships\n 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 2.667798\n ...\n 1.000\n 1.1250 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 \n 4\n survivorsofabuse\n 9p2gbc\n [0, 5]\n October is Domestic Violence Awareness Month
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a...\n 1421\n 1\n 0.8\n 1539809005\n 24\n
7.554238\n ...\n 1.000\n 1.1250\n 1.0\n 1.77644\n
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