

Final Project Report

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Overview:

In this project, machine learning techniques are used to predict the eye state of a human being; whether if his eyes are open or closed. Brain electrical signal are recorded using a biomedical device called: EEG which stands for “Electroencephalography”.

Electroencephalography (EEG), as shown in figure1, is an electrophysiological monitoring method to record electrical activity of the brain. It is typically noninvasive, with the electrodes placed along the scalp, although invasive electrodes are sometimes used in specific applications.

EEG measures voltage fluctuations resulting from ionic current within the neurons of the brain. In clinical contexts, EEG refers to the recording of the brain's spontaneous electrical activity over a period of time, as recorded from multiple electrodes placed on the scalp. Diagnostic applications generally focus on the spectral content of EEG, that is, the type of neural oscillations (popularly called "brain waves") that can be observed in EEG signals as shown in figure 1.

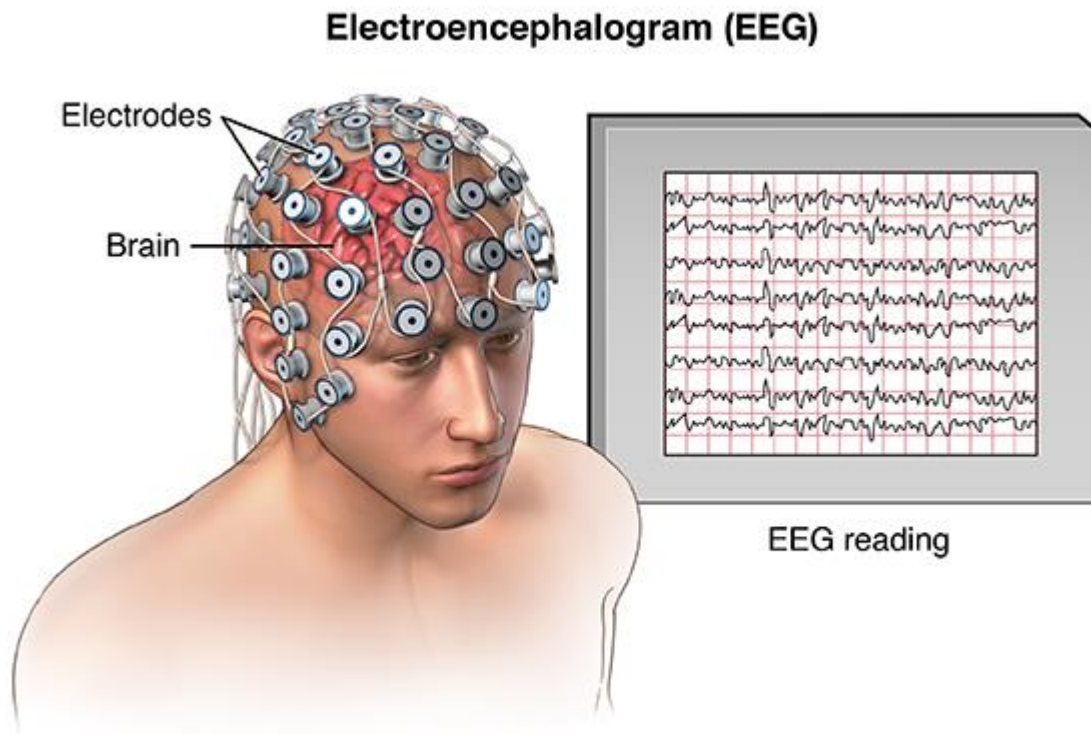


Figure (1)

After obtaining the dataset from this device, signal processing techniques are applied then machine learning to predict the state of the human eye. Various machine learning algorithms are used with various accuracies predictions.

Dataset:

The data set consists of 14 EEG values and a value indicating the eye state. The following table illustrates and gives an insight on the dataset that we have:

Data Set Characteristics:	Multivariate, Sequential, Time-Series	Number of Instances:	14980	Area:	Life
Attribute Characteristics:	Integer, Real	Number of Attributes:	15	Date Donated	2013-06-10
Associated Tasks:	Classification	Missing Values?	N/A	Number of Web Hits:	41231

This dataset is preprocessed and ready for classification. So, we *do not need any preprocessing or feature extraction*. All data is from one continuous EEG measurement with the Emotiv EEG Neuroheadset shown in figure 2. The duration of the measurement was 117 seconds. The eye state was detected via a camera during the EEG measurement and added later manually to the file after analyzing the video frames. '1' indicates the eye-closed and '0' the eye-open state. All values are in chronological order with the first measured value at the top of the data.



Figure (2)

Dataset contains 14980 samples, 14 features*, and one binary label. Data is divided 70% for training and 30% for testing.

Feature standardization is used using Sikitlearn library in python to scale and normalize the feature samples. Then the data is fed to the various model predictions that are used.

*features are the signals from the 14 electrodes: 'AF3','F7','F3','FC5','T7','P7','O1','O2','P8','T8','FC6','F4','F8','AF4'

Prediction Models:

So many algorithms are used in machine learning as a prediction models. To pick a good model that fits your data is not an easy task. In this project, we have tried different models, and obtained different accuracies.

- **Supported Vector Machine (SVM)**

The first model that we used is Supported Vector Machine (SVM) with radial basis function, with the following parameters: kernel='rbf', random_state=0, C=1.

- **Decision Tree**

The second model that we used is Supported Vector Machine (SVM) with radial basis function, with the following parameters: criterion='entropy', max_depth=40, random_state=0.

- **Neural Network**

The third model that we used is NN, but it is discarded from the code lines because of the very bad accuracy even with hyper parameters variations. This model gave a very bad accuracy after several tries tuning the parameters.

- **Random Forests**

The fourth model that we used is Random Forests, which gave the higher accuracy of all the previous models. The used parameters are: n_estimators=1000, random_state=0

Accuracies:

The accuracies for the above prediction models are collected in the following table:

Model	Train Accuracy	Test Accuracy
SVM	100%	54%
Decision Tree	100%	83%
Random Forests	100%	92.4%

As shown in the table, SVM has the lowest accuracy, while Random Forests has the higher accuracy.

Comparison:

To compare my model with other proposed models in literature that used the same dataset, I have searched in literature for a paper until I've found a paper that used the same data. This paper has the title: A First Step towards Eye State Prediction Using EEG by Oliver Rosler and David Suendermann. They have used various models to estimate the best accuracy. I have scored almost the same accuracy as them. As seen in figure 3 (from the paper):

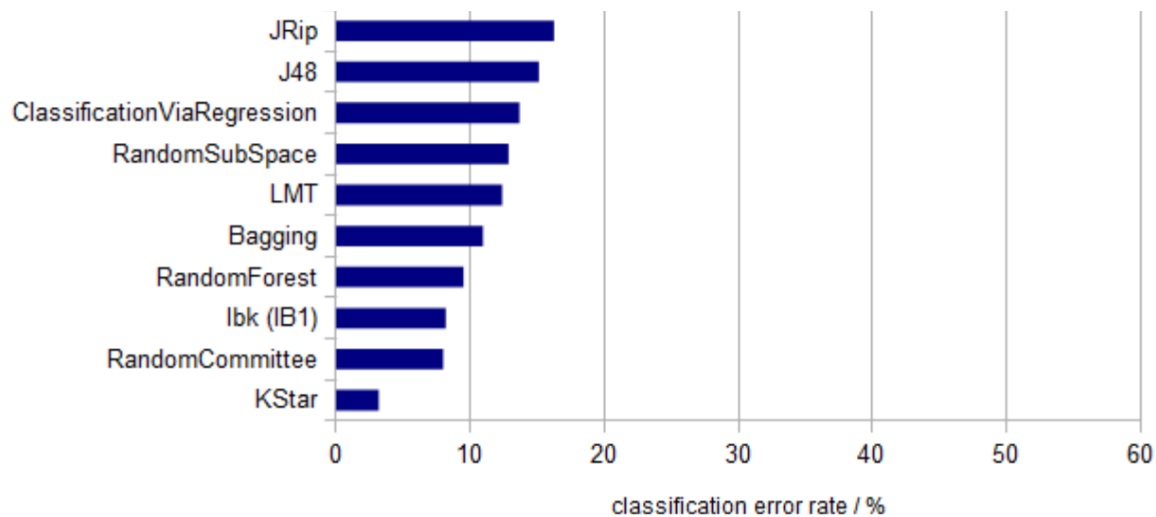


Figure (3)

Conclusion and Results:

According to the random forests, the feature importances are represented in the following figure 4:

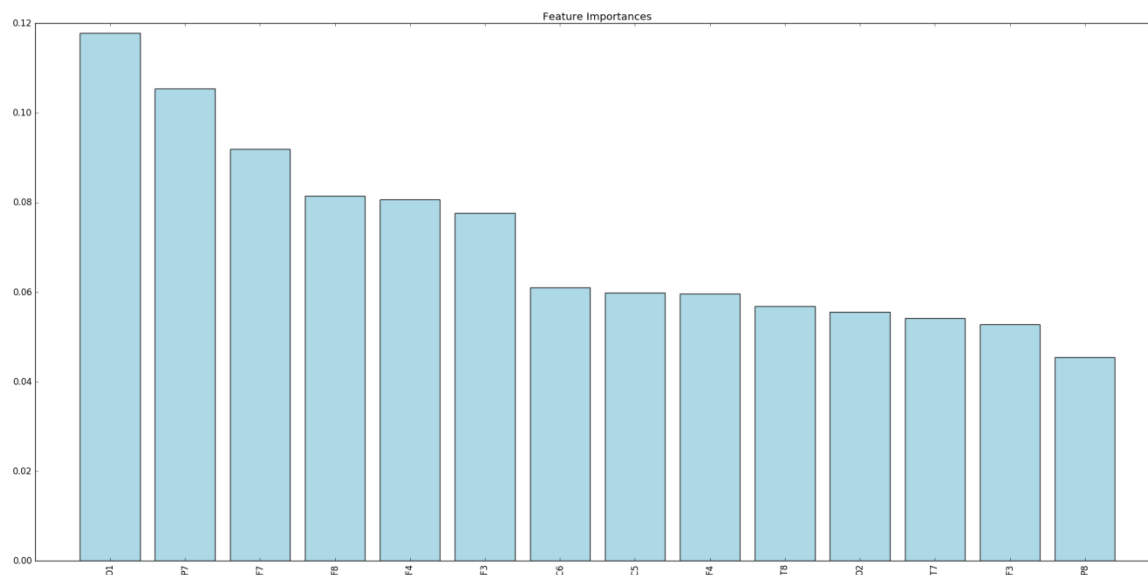


Figure (4)

In conclusion, our model fits the data as the literature paper. We got the same accuracy with random forests,

Future Discussion:

In the future, we may record more signals from more electrodes than those 14 electrodes. We can also try to use deep network or deep learning techniques. Also we can reprocess the feature signals after extracting it from the electrodes and sample it with different sampling than this dataset.

One important thing is to find an application for this eye state. We can use it in automotive cars to detect the human eye if it closed then it stops the car, because sometimes when people driving they fall asleep and this could cause serious accidents.

We can also use this technique for the locked-in syndrome patients. Those people can't speak or move anything but their eyes. Somehow we can detect their eyes if it is open or closed and relate this eye state to a mobile phone for emergencies or sending messages after a certain sequence of open/closed eyes which is at the end a certain sequence of binary data.