

PREDICTING EPILEPTIC SEIZURES FROM EEG DATA COMPUTER SCIENCE AND ENGINEERING, IIT KANPUR ARNAB GHOSH , PROF. AMITABHA MUKERJEE }

OBJECTIVES

- 1. To predict **preictal**(just before seizure) and interictal(interim periods between seizures) phases from the EEG data.
- 2. Try to figure out the patterns in the data that correspond to interictal and preictal phases.
- 3. Use the observed patterns to develop an algorithm to predict seizures based on the EEG data.
- 4. Try to look at the cognitive aspect of the experts how they perceive the data and draw the inferences about the phase of seizure .

ALGORITHM 1

Data: The EEG data

Result: Predict 1(Seizure) or 0(Otherwise) for all channels do

Compute the variance and the

correlation between the channels end

use the Correlation Coefficients to build the Predictor Matrix

while learning from the training data do for each subject do

Learn a set of decision trees (Forest) using the Predictor Matrix.

end

end

while A new EEG sample is given do **for** all the decision trees in the subjects Forest **do**

Prediction of all trees on this sample end

Output the majority of all Predictions end

Algorithm 1: Using Random Forests

REFERENCES

- [1] Stead SM et al Howbert JJ, Patterson EE. Forecasting seizures in dogs with naturally occurring epilepsy. 13(21):1–21, March 2014.
- [2] Chicharro D Andrzejak RG. Seizure prediction: Any better than chance? pages 1–12, July 2009.

MOTIVATION

Epilepsy is the disorder where an aggregate of Brain Cells start firing synchronously leading to abnormally high levels of brain activity and an eventual seizure.

The ability to predict the occurrence of a seizure is an important problem because it would allow eplileptic people lead more normal lives. If we could warn an epileptic patient of a possible seizure beforehand then appropriate medication will help in combatting it , at the same time reducing the side effects of these chemical bound medications.

ALGORITHM 2

Data : The EEG data Result : Predict 1(Seizure) or 0(Otherwise)
while learning from the training data do
for each subject do
for each data sample X(16*239766) do
Obtain a Random Projection
Matrix RP(239766*2400);
Compute the inner product of
$ $ $RP*X$
Pred= SVM with Gaussian
Kernel to train the classifier.
end
end
end
while A new EEG sample is given do
use RP to preject the data into the low
dimensional space
use Pred matrix and data to compute
the Prediction
end
Algorithm 2: Using Support Vector Machines

The methods we tried to apply proved good

enough results but it is still far from state of the

art. We have to look at slightly different mecha-

nisms trying to exploit the time series component

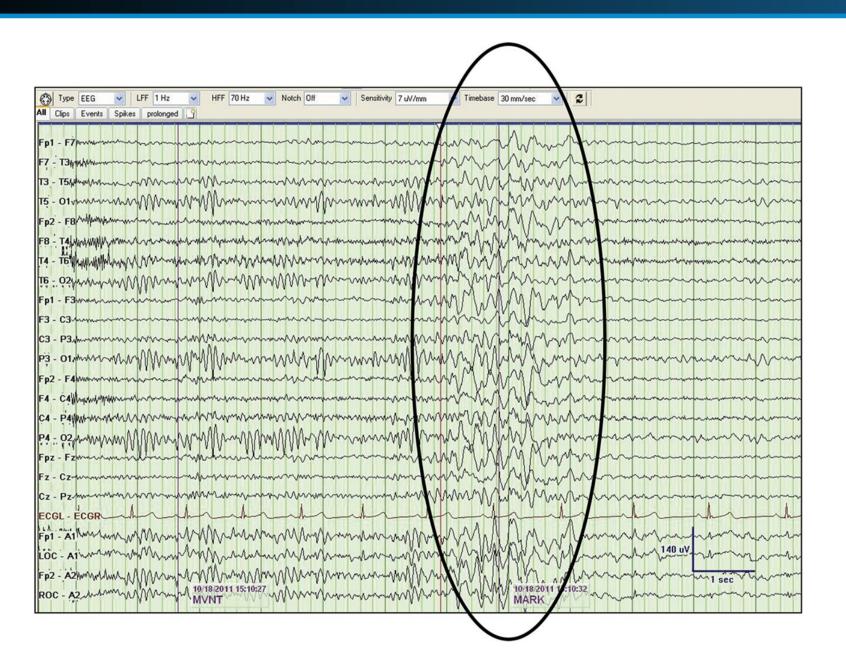
of the EEG data.

Observation 2 Not all channels are equally excited in the case of a seizure.

This is because a seizure generally doesn't impact the whole of the brain but rather concentrated around the Hippocampus, hence we cannot treat the data for all the channels as equivalent . Hence we find out the correlation among the various channels and use it as the Predictor Matrix for the Training of the Decision Trees of the Forest.

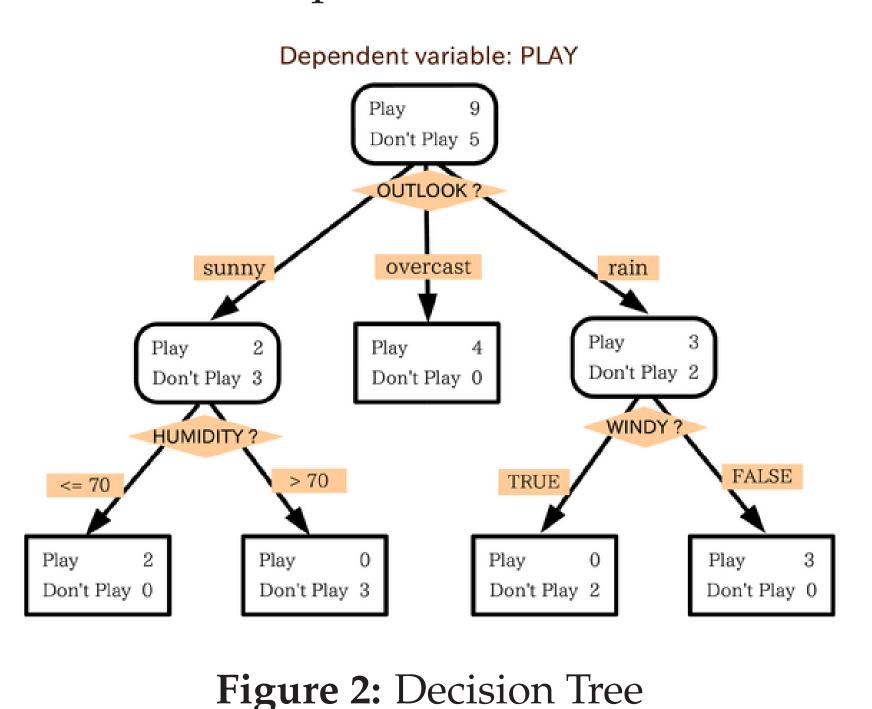
CONCLUSION AND FUTURE RESEARCH If experts well versed in this field can be consulted and understand their perception of the data detection of anomalies using their instinct, we could try some more fancy neural networks and deep learning techniques to solve this problem.

OBSERVATIONS

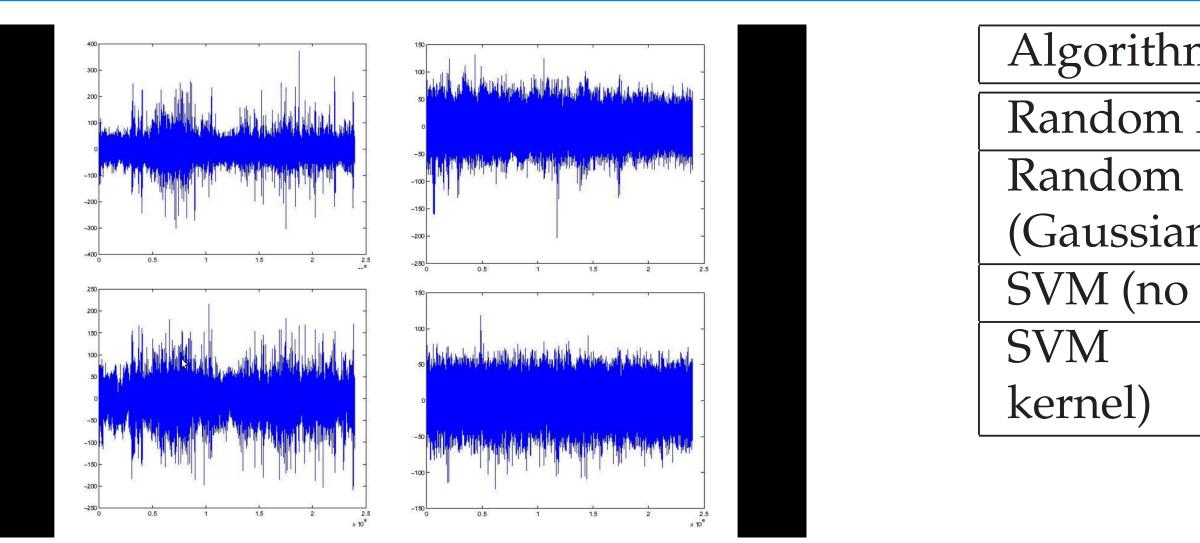


EEG data

Observation 1 Some of the channels , particularly those near the Hippocampus face excessive amount of brain activity. Generally it is seen that during a seizure, the signals of the brain work in a constructive interference pattern and reaches very high levels, ultimately culminating in a seizure. We used this observation to model the SVM and to tune the parameters for better prediction.



RESULTS



CONTACT INFORMATION

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Figure 1: How experts perceive epileptic behavior from

m	F1 Score
Forest	61%
Forest	65%
n Kernel)	
kernel)	70%
(Gaussian	72%