

# Epilepsy Detection By Processing Of Eeg Signals Using My-DaQ And MyRIO

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**Abstract-** Epileptic seizure or epilepsy commonly known as fits have been occurring one in a 100 in day to day life and the causes to this are many. In spite of so many causes it appears physically as much as most of the time to a lot of people. The overview of the work presented over here is that the occurrence of the epilepsy to appear physically can be treated or controlled before in hand. There are various methods to do that. This detection before in hand itself is done by various procedures which includes the methods of capturing and monitoring the Electroencephalograph (EEG) signals from the patients, separating the signals with respect to their frequencies and then comparing it with the threshold with respect to their amplitudes. Crossing the threshold indicates the occurrence or initiation of the Epilepsy and thus is beforehand itself indicated by the occurrence of an alarm. This detection here is being performed by using MYRIO and MyDAQ wherein using Labview block diagram is dumped, the EEG signal samples are given as input and the objective is achieved.

**Keywords-** Epilepsy, EEG, threshold, simulation, Labview, MyRIO, MyDaQ

## I. INTRODUCTION

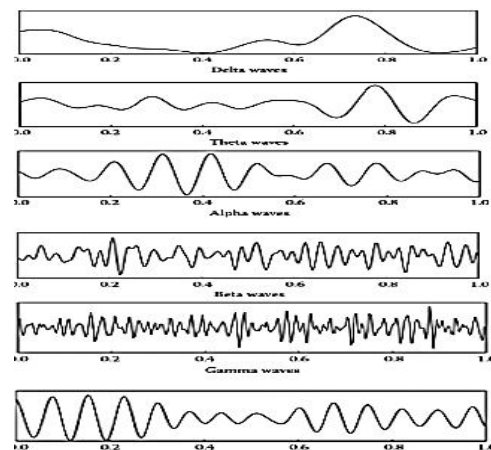
### 1.1. Introduction to EEG Signals:

The recording of the electrical activity of the brain along the scalp which is a result of current obtained due to the firing of neurons in the brain coz of fluctuation of voltage is known as Electroencephalography (EEG). The brain gives a response to the stimulus and thus the measurement of the response to the stimulus is known as Event-Related Potential (ERP). The stimulus may be sensory, motor or cognitive in nature. Various patterns of EEG are recorded in the table below as follows:

**Table 1. EEG wave patterns and their frequency ranges**

Sl. No	Wave Patterns	Frequency (In Hz)
1	Delta	4
2	Theta	4-7
3	Alpha	7-14
4	Beta	15-30
5	Gamma	30-100
6	Mu	8-13

Some brain wave rhythms and waves whose frequency ranges as indicated above are being diagrammatically represented as below.

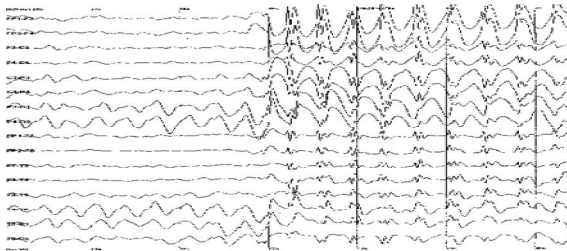


**Fig 1. Various Rhythmic patterns in the human brain**

### 1.2. Epilepsy:

Epilepsy is completely a chronic disorder which is always recurrent and is always provoked by the unprovoked seizures. The people who have fits or epilepsy have symptoms more than one type of seizure and they most of the times of have other neurological problems as well. Epilepsy may also be said as a spectrum condition with a wide range of seizure types and the control varies from person to person. Epileptic seizures signals characterized from small range low signals to vigorous shaking. The cause of the epilepsy is unknown though most of the defects are birth defects and sometimes they also occur due to the electrolytic imbalance in between the neurons. The most common type of the seizures is known as convulsive and its constitute about 60% of the total seizures

that are present. Among which 1/3rd of them occur start occurring due to its affect in both the hemispheres of the brain. 2/3rd of all the seizures are due to the partial affects that are being occurred and then completely affect the person. The rest 40% of the seizures are non-convulsive. About 6% of the epilepsies that are caused in the world are due to the seizures that are being triggered by the specific events which are also known as reflex seizures. These reflex seizures always occur as a result of a specific stimulus.



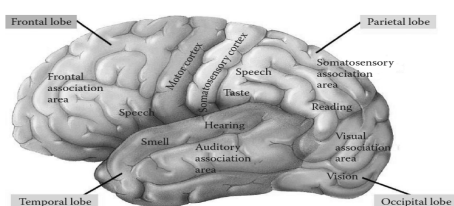
**Fig 2. EEG of a person with childhood absence epilepsy showing a seizure**

**1.3. Source of Neural Activities:**

The home for neural activities for the origin of EEG signal may be human brain and the neuron. The most fundamental cell in neuropsychology is neuron from which the firing occurs and thus EEG come into picture. Thus the following has been presented in the following sections below.

**1.3.1. Human Brain:**

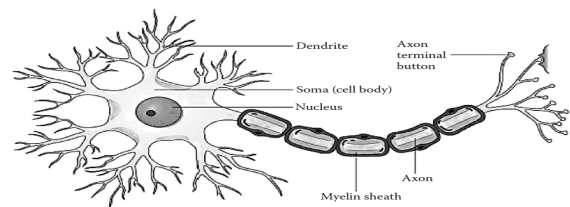
The human brain is divided into 3 parts: Cerebrum, Cerebellum and Medulla Oblongata. The major concentrated parts for EEG reference are Cerebrum and Cerebellum. Complex body movements including co-ordination and muscle tone modulation are controlled by cerebellum. The cerebellum is again divided into 6 parts. They are frontal lobe, temporal lobe, insular lobe, limbic lobe, occipital lobe and parietal lobe. Pain, taste sensations and problem-solving activities are being concentrated in the parietal lobe. Hearing and Memory are associated with temporal lobe. Vision-related tasks are being taken care by occipital lobe. The frontal lobe is associated with problem solving, emotions, movement and speech.



**Fig 3. Functional diagram of human brain**

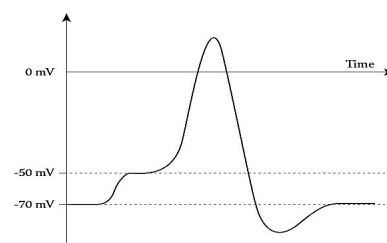
Somatosensory cortex and premotor cortex is a location where a primary motor cortex lies. Different areas of the primary motor cortex control different movements of the body. Upper part of the primary motor cortex controls shoulders, elbow, wrists, fingers and thumb. The larger area of the primary motor cortex is being occupied by the facial movements such as eye, lips, face, neck, tongue, jaw and swallowing. Any movements occurring in the body gives rise to some specific signals from primary motor cortex. Thus for the assessment of motor related brain activity, the EEG signal recording should be originated from the appropriate location in the spatial domain for the correctness reading of each electrode.

**1.3.2. Neuron:**



**Fig 4. Structure of a neuron**

The main components of the Central Nervous System (CNS) are nerve cells and glia cells. Every neuron has cell bodies, dendrites and axons. The function of the nerve cells is to respond to stimuli and to transmit the information over long distances. The sending and transmitting of an electrical signal occurs through an axon which is a long cylindrical tube. The proteins to the end of the cell is being delivered by axon's transport system. The resting membrane potential of neuron is about is about -70 to 60mV.



**Fig 5. Action Potential of a Neuron**

**1.4. Measurement of EEG Signals:**

The electric potential generated by a single neuron is very much insufficient to be captured up by EEG and thus a collective data of millions of neurons that are synchronously working with similar spatial orientation is collected and then the EEG is analyzed. The cognitive state changes and the sleep states of the neurons are being analyzed by the EEG. EEG data can be acquired using 2 types of instruments: Digital and

Analog EEG instruments. Analog Instruments consists of an amplifier, a writing device and a galvanometer where the output of the amplifier is always being controlled by low frequency and high frequency filters and sensitivity controls. Ex: A sensitivity of  $10\mu\text{V}/\text{mm}$  means that a signal with amplitude of  $100\mu\text{V}$  will produce 1cm deflection on the vertical scale. The analog waveform is being converted to digital values and thus this process is known as Analog to Digital Conversion where the continuous signal is being quantized according to a particular interval and quantization values are being given to the signal with respect to its amplitude.

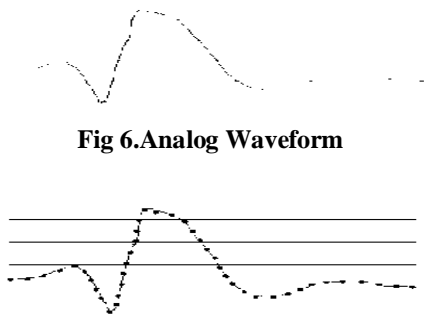


Fig 6. Analog Waveform



Fig 7. Sampling rate of 240Hz

The sampled waveforms can be then displayed as waveforms after storing and manipulating the values that are being stored. The rate at which the waveform data is being sampled is known as Sampling Rate for which minimum acceptable range is 2.5 times greater than the highest frequency as most digital EEG systems support sampling rate of 240Hz. Some recordings from brain directly have a frequency of nearly 200Hz. Some digital systems have optional sampling rates of 480Hz which are then being converted using analog to digital converter after being passed on with the anti-aliasing filter.

#### 1.5. Sensors / Electrodes:

The electrodes can be categorized Based on

- Circuitry design
- Conduction
- Communication

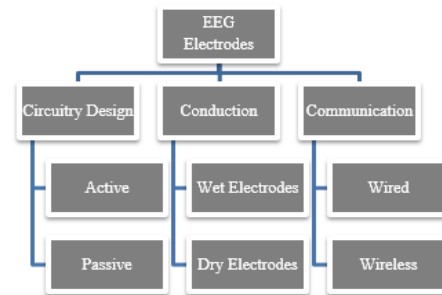


Fig 8. Classification of Electrodes

#### Based on Circuitry Design:

Electrodes are classified based on whether the circuitry is inbuilt within them or not.

- **Active Electrodes:** They have inbuilt circuitry where the current is being amplified due to which the signal quality received is greatly improved and also the skin preparation and conductive pasting required by passive electrodes can be avoided.
- **Passive Electrodes:** They don't have inbuilt circuitry and thus the results are less efficient than the active ones

#### Based On Conduction:

- **Wet Electrodes:** Wet electrodes rely on electrolytic gel to penetrate hair wherein the gel also acts as a medium to fill up the air gaps in between the sensors and the electrode and provide a clean conductive path
- **Dry Electrodes:** Dry electrodes directly comes in contact with the skin wherein the electrolytic gel is not necessary at all. Here since the sensors have to be in their place intact and thus to avoid the noises and artifacts mechanics and electronics comes into picture. **Ex:** Quick 20.

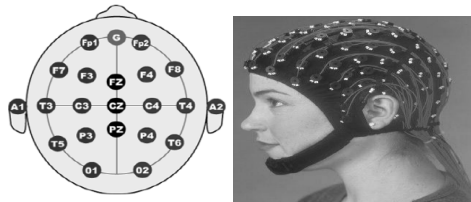
#### Based on Communication:

- **Wired Electrodes:** The electrodes are being attached with wires and are being connected to the analyzing center and thus the EEG signals that are being transferred to the processing center are processed. **Ex:** PO8
- **Wireless Electrodes:** The electrodes are not being attached by any wire and thus the information collected is being transmitted all wirelessly. **Ex:** Quick 20

#### 1.6. Positioning of Electrodes:

The positioning of electrodes is one of the important aspects in the EEG signal capturing and detection. For this purpose the 10-20 Method is being used wherein the electrodes are being placed along the scalp of the head. The

10-20 method is based on the relationship between location of the electrode and underlying area of the cerebral cortex. The positioning of the electrodes along the scalp is shown as follows in the figure depicted below.



**Fig 9. International 10-20 Placement System**

The letters used indicate:

F for Frontal Lobe  
 T for Temporal Lobe  
 C for Central Lobe  
 P for Parietal Lobe  
 O for Occipital Lobe

“Z” indicate that an electrode is being indicated on the midline.

#### 1.7. Abnormal EEG Patterns:

Abnormality is being indicated by certain variations in the EEG signals that are being generated by the brain. The reasons for the following may be due to the following as such as increase in the abnormal EEG patterns, decrease in the abnormal EEG patterns or sudden disappearance of the EEG patterns and thus the non-specific abnormalities in the EEG signals can be categorized into 3 categories. They are as follows:

- Intermittent slow wave abnormalities that are widespread and are usually associates with brain dysfunction and are seen in delta wave range.
- Persistent bilateral EEG, most of the times usually associated with the impaired conscious reactions of the cerebral hemisphere.
- EEG associated with disturbance of cerebral hemisphere focused.

A burst-type signal is being associated with the first-category signal which is reduced/ attenuated by alerting the particular individual by eye-opening and then it can also be fastened or accentuated by drowsiness, closing of eyes, etc. The maximum amplitude is being usually observed in the frontal regions and is influenced by age. This is most prominently observed in the occipital and posterior head region in children. It is also observed as the most prominent

combinations of the pathological processes including system metabolic and toxic disturbances in the frontal regions.

In the second category, i.e. bilateral persistent EEG, the occurrences and the processes among different stages of conscious, impaired, purposeful responsiveness are totally non-specific and the mechanisms responsible for the following above occurrences are only partially understood.

In the third category of classification, i.e. focal persistent EEG, the abnormalities are noticed in the form of disappearances or distortions of the normal patterns occurring in the EEG. Increase or appearances of all abnormal patterns but such kind of changes are always seen at the cerebral cortex. The focal distortion of the normal rhythmic patterns results in an asymmetry in frequency, reactivity or amplitude of the rhythm.

The focal abnormalities may also be the reason for the rhythmic delta waves being produced and thus this is the most reliable findings in the focal cerebral disturbance. These disturbances resulting in the generation of the delta waves are more polymorphic and non-rhythmic and are always less-reactive but are more reliable.

#### 1.8. Objectives:

The Epileptic Seizure more oftenly called as FITS occurs at any time because of some of the abnormalities that occur within the human brain. The main reason that the epilepsy occurs is due to the abnormal sparking of the neurons which inturn is caused by the irregular activities within the brain. Thus the person may not have medical aid at that particular instant of time and may collapse. Since occurrence of epilepsy gives some instincts or appearance symbols some few hours before its actual physical appearance, capturing its initial signals of start of it will be notified by the comparison with the threshold set by the hardware and thus this gives an indication of its occurrence before its actual physical appearance by the display and buzzer indication and the patient can survive.

## II. LITERATURE SURVEY

A common chronic neurological disorder, marked by seizures is known as Epilepsy which is due to excessive discharge of cortical cells from the brain. The epileptic seizure is undetectable and cannot be completely cured but it can be controlled through proper medicinal drugs. The detection of seizures is highly needed since they are unpredictable and occur very rarely. Here, some of the previous methods that

were used for the automatic detection of epileptic seizure in EEG signals are discussed.

Logesparan & Rodriguez-Villegas in the year 2011 proposed an algorithm which is based on data selection which was based on phase congruency and determining interictal spikes from background EEG. A modification of the traditional phase congruency was being involved in the approach where dynamic estimate of the muscle activity along the scalp is included. The performance was greatly increased by 80% even if the data was reduced by more than 50%.

A Novel approach by Guo in 2010 for the automatic epilepsy seizure detection that used the line length features based on the decomposition of the wavelet transform. The EEG signals for detecting the presence and absence of the seizure was being done by the Artificial Neural Networks. The preprocessing of the database was being done by removing the artifacts. In the year 2011, automatic feature extraction was being done from the original feature database in epileptic classification of EEG signals and this was done by applying Genetic Programming (GP).

Hierarchical EEG classification system which may be useful in detecting epileptic seizures was proposed by Wang in the year 2011. The proposed method includes 3 steps:

- Representing the original EEG signals by wavelet coefficient of packets and its feature extraction by best entropy available
- k-Nearest neighbor classifier and Cross-Validation (CV) approach for Hierarchical Knowledge Based (HKB) construction.
- Accuracy Computation and rate rejection to be calculated.

Proposed method was Channel selection which was done by Chang in the year 2010 which reduces the channels from 22 to less than 6. Thus computational load can be reduced. By doing so 70% increase in efficiency in 3-channel cases of EEG database.

In the year 2010, Liang, in order to detect the seizures a systematic evaluator approach was used. With a combination of spectrum analysis and complexity analysis, the evaluation can be done. Genetic Algorithms and Principal Component Analysis (PCA) were applied to linear and non-linear methods.

A method for establishment for differences in power spectra between EEG channels and successive time intervals was being introduced by Lewis in the year 2010. The differences among the time intervals during seizures can be

minimized by characteristic evolution of seizures. The combination of Discrete Finite Automata (DFA) and Spectral analysis is very much useful in seizure identification.

With an objective of evaluation of the estimation of the lateralization of epileptic EEGs and preprocessing effect on synchronization analysis, a synchronization analysis on EEG was carried out by Verez Perez in the year 2011. Preprocessing includes cancellation of noise by Wavelet Denoising (WD), Blind Source Separation (BSS) and classification. This method has an efficiency of 90% in locating the epileptic zone.

A Discrete harmony search proposed by Gandhi in the year 2012 which is for detecting epileptic seizure activity from background EEG data recorded from normal and epileptic subjects. Discrete Wavelet Transform (DWT) with entropy, standard deviation, mean, kurtosis energy and entropy estimation was included in the approach and later followed by Probabilistic Neural Network (PNN). This is highly in demand and thus has 100% efficiency.

Guo in the year 2010 presented a method of automatic epileptic seizure detection which uses entropy features derived from Multi Wavelet Transform (MWT) combined with an ANN to detect the presence of seizure but increases the cost of computation.

### III. METHODOLOGY

Initially in the first phase, the complete analogy of epilepsy (Fits) is being gone through and then the concept of the EEG signal that is being generated by the brain is being studied. The concept of the EEG signal generation, capturing and processing is also being studied but here processing of the EEG signal is not within the scope of this current implementation and thus it will be studied based on the upper surface. The voltage levels and the action potentials of the neurons are being studied during their normal activity and abnormal activity.

In the second phase, the components required for the current implementation such as the electrodes, types of electrodes needed, its specifications are gone through. The interfacing devices are also being studied which are necessarily required for the signal capturing. Analog to Digital Conversion specifications and various factors affecting the conversions are being studied. Comparators are also being taken into considerations and thus its specifications are chalked out. Various other components utilized are also being studied in this phase.

In the third phase, plans are being chalked out in order what way to implement the current implementation. In this phase, initially the block diagrams are being implemented according to the components necessary step by step. Then combining all the blocks together and assigning the inputs and outputs are being taken out.

In the fourth phase, all the components are being practically implemented according to that as per chalked out in the third phase as per the block diagram. The components are properly connected and then the connections are being analysed whether they are connected as per the circuit required. The inputs are properly connected and the outputs are properly taken out and analysed in the next phase of the current implementation.

In the fifth phase, the results are being obtained by applying various inputs at the input side and thus the outputs are being analysed based on the applied inputs. The output is checked and verified as to confirm whether the expected output is obtained or not. The conclusion is being drawn based on the results obtained and thus analyzing them. Also based on the implementation done, the advantages and disadvantages of the current implementation can be drawn and may be notified. The following figure below depicts the implementation process of the current implementation.

**IV. LABVIEW SIMULATION**

The EEG signal extraction, processing and reporting can be done by using various methods. But a particular method has to be followed for the acquisition of EEG signals. Thus the following method or stages are as mentioned below are as done.

**A. EEG Extraction:**

The EEG signals captured from some about wireless EEG cap that are being placed on the patient’s scalp is being captured as in the file formatted as “.edf” extension. The examples for wireless EEG sensors used are EMOTIV EPOC, EMOTIV INSIGHT. The following figure below depicts the actual wireless EEG sensors.



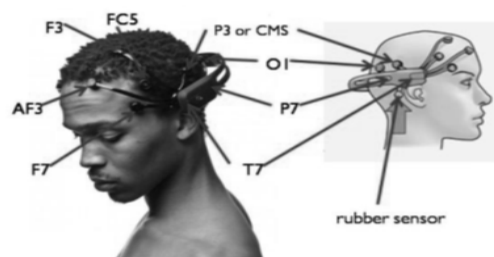
**Fig 10. EmotivEpoC EEG Headset**

This then can be later used in processing. The extracted and transmitted signals also contain the signals which may be unnecessary in actual processing. The unnecessary signals need to be filtered out since the EEG signals are in the range of 0.5Hz to 100Hz. The divisions of the frequencies are shown as follows.

Brainwave Type	Frequency Range	Mental states and conditions
Delta	.5Hz to 3Hz	Deep, dreamless sleep
Theta	4Hz to 7Hz	Fantasy, imaginary dream
Alpha	8Hz to 12Hz	Relaxed but not drowsy
Low Beta	13Hz to 15Hz	Formerly SMR
Midrange Beta	16Hz to 20Hz	Thinking, aware of self & surroundings
High Beta	21Hz to 30Hz	Alertness, agitation
Gamma	30Hz to 100Hz	Motor Functions, higher mental activity

**Table 1. Table 1. Frequency Ranges of EEG Signals**

A clinical routine checkup of an epileptic seizure or epilepsy involves a EEG signal recording of nearly 20-30 minutes and this recording usually involves the capturing the signals from the electrodes attached to the scalp. The EEG amplifier such as EMOTIV INSIGHT and the electrodes to the scalp are used to capture the signals. The raw data captured from the electrodes attached to the scalp of the person are sent to the computer wirelessly, i.e. Wi-Fi or Bluetooth and the file is then saved with an extension “.edf”. The electrodes are attached as shown in the figure below.



**Fig 11. EEG Electrodes attached and EEG Measurement points using EMOTIV-EPOC system**

EEG signal can be very easily simulated in the Lab VIEW using the biomedical toolkit. The signal is always analyzed to determine the current human state. The signals are simulated using EEG.

The signals that are being captured from the electrodes include the noise and external signals from the environment. The disturbance in the signals that are being separated by the use of filters. Various EEG filters are available within the Lab VIEW such as IIR and FIR filters. A linear phase response is being provided by FIR filters and a

non-linear phase response is being provided by IIR filters. Thus FIR filters can be used in applications where phase information is necessary and IIR filters are being used where phase information is not necessary.

The major difference in between the FIR and IIR filter is the impulse response among them where it is finite for FIR filters and infinite for IIR filters. FIR filters always typically require more number of multiplications and summations than compared to IIR filters for the same filtering performance. Since, certain computer architectures are very much suited for the digital computation of the given architectures such as for digital signal processors (DSP's) which are frequently very much better suited for the FIR filtering. The computation speed of FIR filter is very much higher than the IIR filter

IIR filter always provide very high co-efficients values as compared to the FIR filters and this with a lot of co-efficients being available, the computations are easily possible and the computational speed also increases and thus is much more efficient than an IIR filter.

The captured “.edf” file is being read using the Lab VIEW biomedical module. The extracted EEG signals are being filtered out using the specific filter that is being designed and thus the specific EEG rhythm is obtained. The specific IIR butterworth bandpass filter. The Delta signal is being obtained by setting the lower frequency of bandpass filter as 0.5Hz and the higher frequency as 3.9Hz. Similarly to obtain the Theta waves the frequency boundaries of the bandpass filter has to be set to 4.1Hz and 6.9Hz. The Alpha waves are got passed through the bandpass filter having the passband being set as 8Hz and 12Hz. The frequency band for the bandpass filter that are being set to 13Hz and 30Hz to obtain the Beta Frequency. The signal boundary frequencies for the passband in order to obtain the Gamma Signals are 30Hz and 100Hz. The Amplitudes and the frequencies are being extracted using single tone measurement module. In order to repeat this for N times a loop is being used. The following diagram below depicts the block diagram for signal filtering.

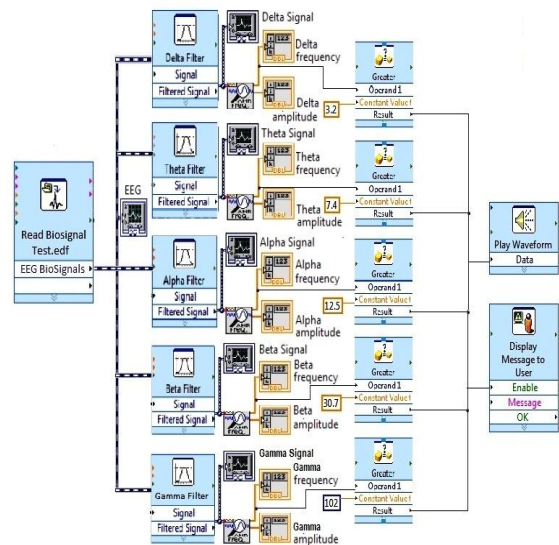


Fig 12. Block Diagram for Signal Filtering

The filters used here is the Butterworth filter. Adjusting the band frequency boundaries is as follows shown in the figure below:

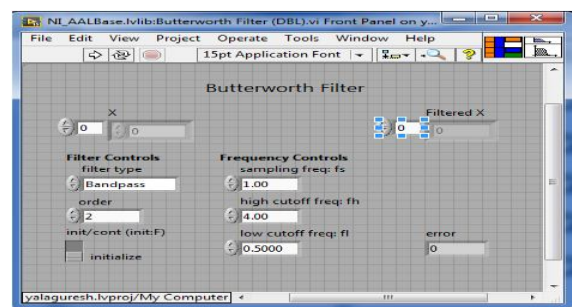


Fig 13. Filter Parameters for Delta

Similarly, the parameters for all the other 4 EEG signals are set. Plotting the block diagram gives the front panel view as shown below.

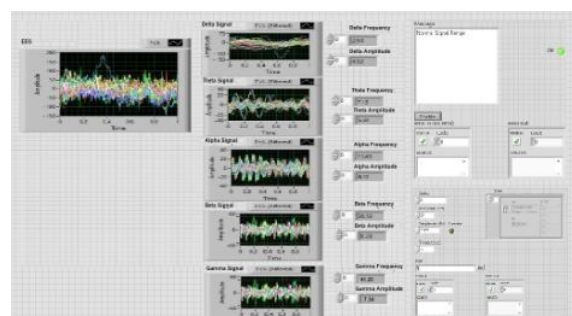


Fig 14. Front Panel View for the Block Diagram





being transmitted wirelessly to the MyRIO module and is then received.

The MyRIO module is being programmed/dumped with the block diagram being drawn using NI LabVIEW software. This dumping will internally map itself to the pins of ARM Cortex M8 in the MyRIO which in turn is being interfaced with the Data transfer instructions with the LCD module and the buzzer. Apart from the interfacing that is being done, the developed code is being dumped on the FPGA board. The block diagram depicting the internal structure of the LabVIEW MyDAQ can be shown below as follows.

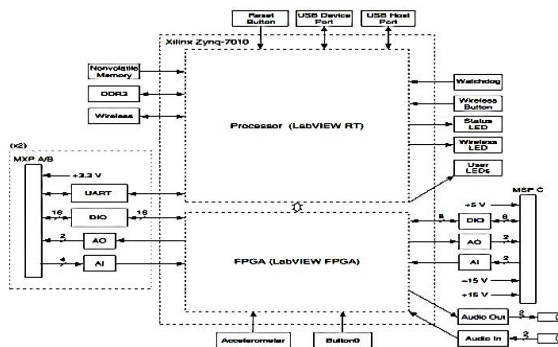


Fig 20. A Complete Block Diagram of the Hardware of NI MyRIO

Choosing Acquire Signals and then choosing the option of Analog Signals with which Voltage Signals has been selected as the option since EEG Signals are voltage signals. Choosing the kind of signal and Clicking on Finish button will move on to the next screen where the parameters such as voltage range for the frequencies in which the EEG signals lie within. The voltage range for the EEG signals are to be set as  $-50\mu V$  to  $50\mu V$  since the maximum voltage for the gamma signals is  $100\mu V$ .

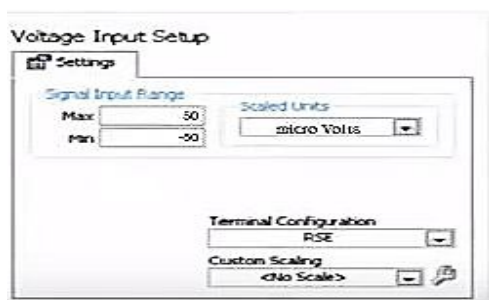


Fig 21. Parameter Value Settings

The timing has to be set for the frequency that has been given a range for and thus the options are being provided for the sampling of the signal that has been captured. Thus 4 options are being provided and since the signal captured is an analog signal and continuously need to be sampled so the

“Continuous Sampling” option is being selected and then click OK.

Once the connection to the graph is given from the data that has been captured and then clicked on the play button in the front panel viewer, then the signals will start appearing on the screen or the waveform chart as shown in the figure below.

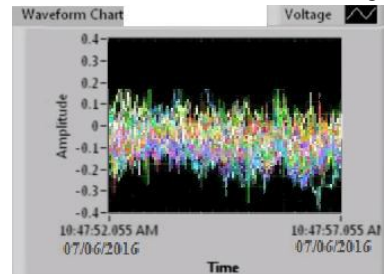


Fig 22. Captured EEG Signals Appearing on the screen

## VI. RESULTS

The .edf file source obtained is then given as the input. The results obtained due to the set threshold are as follows:

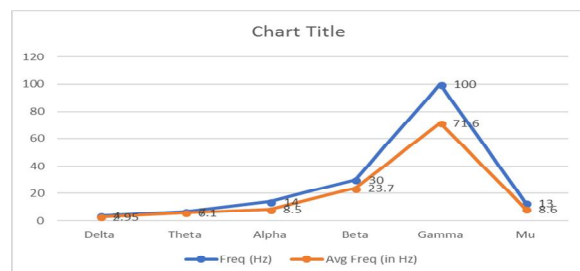
### Case 1: Normal Person

The EEG signal samples were collected from five different normal persons and the results were averaged to obtain the different average wave values.

Table 2. EEG Frequencies obtained in Normal Persons

Wave	Frequency (In Hz)	Limit Lower / Higher	Avg Obtained Freq (in Hz)
Delta	4	0.5Hz / 3.99Hz	2.95
Theta	4-7	4.0Hz / 6.99Hz	6.1
Alpha	7-14	7.0Hz / 13.99Hz	8.5
Beta	15-30	15.0Hz / 29.99Hz	23.7
Gamma	30-100	30.0Hz / 99.99Hz	71.6
Mu	8-13	8.0Hz / 12.99Hz	8.6

A graph is plotted for the obtained average frequency from the EEG signals of normal persons.



**Fig 23. Graph of averaged frequencies obtained from normal people**

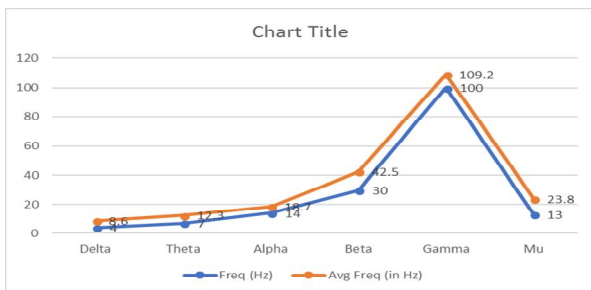
**Case 2: Epileptic Seizure Affected Person.**

Initially a sample of the “.edf” file is obtained from the patient who suffers from Epilepsy. This file is given to the LabVIEW simulation model that is dumped onto a MYRIO board. The frequencies obtained from the epileptic EEG signals are as shown in the table below.

**Table 3. EEG Frequencies obtained in Epileptic Seizure Affected Persons**

Wave	Frequency (In Hz)	Limit Lower / Higher	Avg Obtained Freq (in Hz)
Delta	4	0.5Hz / 3.99Hz	8.6
Theta	4-7	4.0Hz / 6.99Hz	12.3
Alpha	7-14	7.0Hz/13.99Hz	18.7
Beta	15-30	15.0Hz/29.99Hz	42.5
Gamma	30-100	30.0Hz/99.99Hz	109.2
Mu	8-13	8.0Hz/12.99Hz	23.8

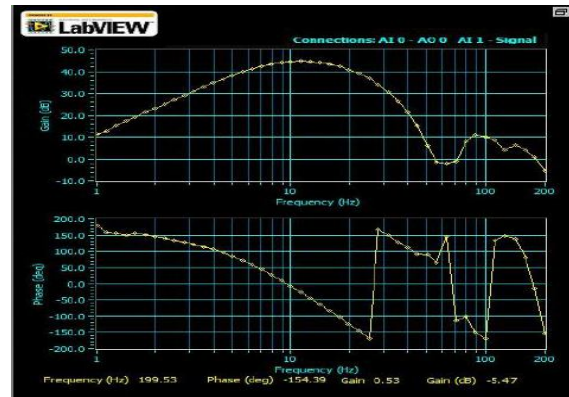
A graph is plotted for the obtained frequency from the EEG signals of Epileptic Seizure affected persons.



**Fig 24. Graph of EEG frequencies obtained from Epileptic Seizure Affected people**

Simultaneously, when the frequency crosses the threshold limit set. The alarms are triggered and the LED is lit in order to notify that the person has been affected with the epileptic seizure.

Separating all the signals finally the complete frequency response graph obtained may be depicted as shown in the figure below.



**Fig 25. The final frequency response graph obtained**

**VII. CONCLUSIONS AND FUTURE SCOPE**

*a. Conclusion:*

The results obtained from the hardware implementation of MYRIO is perfectly obtained with alarms being triggered using buzzer and LED’s lit when the EEG signal frequencies cross the threshold being set. This helps the person to take precautions in timely manner so that further effects of the seizure can be avoided.

*b. Future Scope:*

The future scope of the project lies in the same implementation being deployed using convention method using basic electronic circuit components are planned to be deployed.

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