

King Mongkut's Institute of Technology Ladkrabang  
International College

# **Wireless Brain-Machine Interface System Based on Multi-Channel Visual Flickers**

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## Abstract

Brain-Computer Interface (BCI) enables peoples with disabilities communicate the outer world by using electroencephalogram (EEG) signal. Recently, steady-state visual evoked potential (SSVEP) is employed as a stimulator due to it plays an important role in the response to various visual parameters, i.e., flickering rate (F), intensity (I), and duty cycle (D). The SSVEPs are practical and useful in research because of its excellent signal-to- noise ratio and relative immunity to artifacts. However, another one of the most efficient paradigms, namely, the event-related potential (ERP) based BCI. As a result, the visual stimulation is being recognized to be a most efficient stimulus for a multiple BCI system. In this paper, therefore, we offline investigated the effectiveness of SSVEP and ERP visual flickering paradigms for BCI system in terms of optimizing the parameter of visual simulation. Different experimental conditions were compared to achieve the best performance. Subjects were instructed to fixate LED light source then record associated SSVEP waveform on O1 and O2 channels. The variation in displaying of the presentation stimuli during a task was examined thereby demonstrating the high usability, adaptability and flexibility of the visual stimulator and determine the optimal visual parameters for the subject comfort.



## **Acknowledgements**

I would like to express (whatever feelings I have) to:

- My supervisor
- My second supervisor
- Other researchers
- My family and friends



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# **Chapter 1**

## **Introduction**

### **1.1 Motivation**

Nowadays, The technology is used for the advantages in many fields, for example an ultrasound machine in medical field, a soil and crop sensor in agriculture field, a robotic arm for move the product in industry field and also the several technologies for facility in daily life. At present, one of the technology which is interesting and important is technology to improve quality of life of the disabilities. The invention of equipment which can be interacted by using non-muscular communication channel can help these kinds of people in daily life or in emergency.

For those people with physical or mobile impairment, one of the common problems in daily life is that they cannot completely control an equipment that is designed for ordinary people. Making the equipment for the disabled people to help them can be interacted or communicated by using brain signal instead of muscle that is one of choices to solve their problem. The technology that obtains the brain signal and translates the brain signal to communicate or interact with the external device is called Brain-Computer Interface (BCI). At present, there are three types of BCI, divided by the stimulation : Auditory BCI , Imaginary BCI and Visual BCI. Each type has the different advantages and disadvantages. First is the

auditory BCI that is easy to analyse the brain signal but it has slow interpretation and the sound from this BCI can annoy other people. Next, imaginary BCI is the fastest way to stimulate the brain but the subject should be trained before using this BCI and it has low rate of success. The last one is visual BCI that is the easiest way to obtain the brain signal However, it cannot use in case of the epilepsy person.

Our project will use the visual type of Brain-Computer Interface to control the application for the disabled people. We use brain signal from the subject to control programs or devices. When the user was stimulated by the flicker, the brain will generate waveform and be different pattern from others.

## 1.2 Objective

The objective of this project is to make the user able to communicate or interact with our equipment by using brainwave. This can help the disabled people when they need some emergency or facilities.

Furthermore, our purpose is to study various techniques to record and translate the brainwave to interact with the external devices.

## 1.3 Scope of work

The scope of this project can be summarized as follows:

- Study the theories and techniques of recording EEG waveform.
- Make an equipment which a user can cognitively interact or use the system by brain-wave without using muscle function.

## 1.4 Contributions

In this project, we have developed a control module for users to interact with software by using EEG signal. Our work demonstrates a brand new alternative way to help users in a more convenient way or help immobilize-able users with limited control on their parts of body. We hope that a control module and software that we have developed will provide great advantages and only minor disadvantages to the project.

## 1.5 Procedure

We separate that work into 6 phases.

1. Plan the project
2. Research and study on EEG cognitive
3. Design the equipment
4. Implement the system
5. Doing an experiment
6. Testing program

# **Chapter 2**

## **Problem description and related work**

### **2.1 Problem description**

Nowadays, there are many people who have been born with a disabled body from various reasons. Most of them always have a limitation of their living that be different from ordinary people. At present, one of the main problems of the disabled person is lacking help and ignore from other people. Using knowledge from the scientific and engineering field to make a brain-computer interface machine is the way to help these people to have the living like the ordinary people. For our project, we decide to assist the patient or physically disabled people that can't completely move their body part to interaction without helping from other people such as Paralysis, Spinal cord injured, Stoke and ALS.

### **2.2 Review of related works**

#### **2.2.1 Multitasking with BCI machine[1]**

The multitasking with BCI machine experiment from EPFL Professor Jos del R. Millan. This experiment develops the BCI machine which allows the user to control the robot direction to move by using EEG signal and real-time communicate with other people via the

camera that mount on the robot.

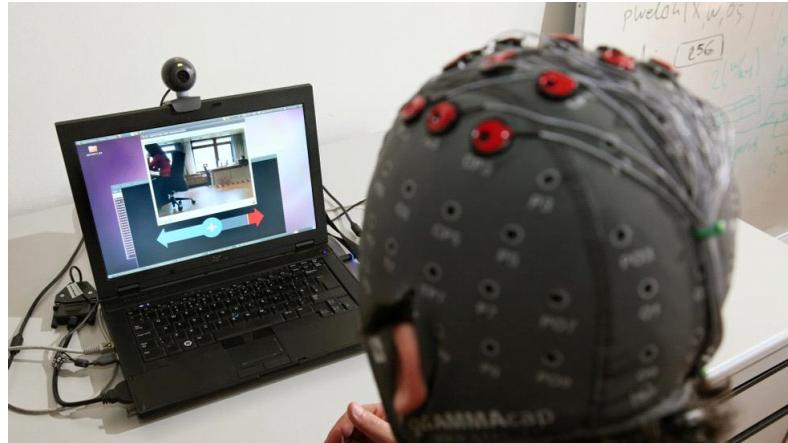


Figure 2.1: The subject control robot use brain wave and communicate via camera at same time

### 2.2.2 Real-time control of a robotic arm using a brain-computer interface[2]

SSVEP control of robotic arm experiment from the Applied Signal Processing in Engineering and Neuroscience Lab (ASPEN Lab) of Old Dominion University.

This Lab is used to develop an application which using EEG headset to control the robotic arm. The subject should gaze on the one of the flickers in front of him for control or command the direction of the robotic arm.

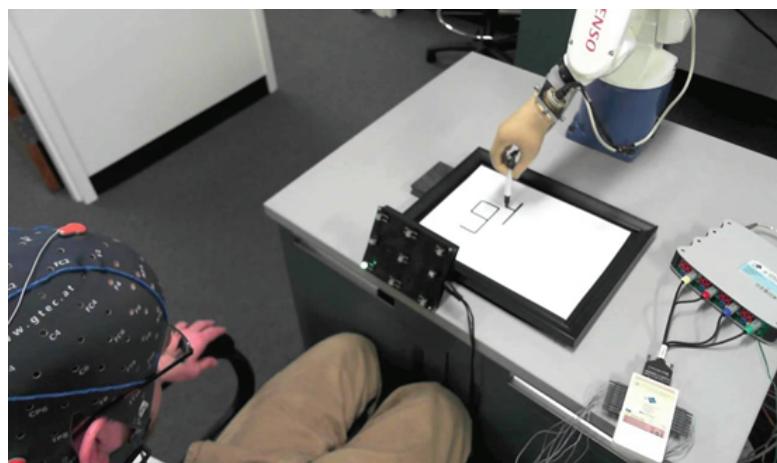


Figure 2.2: The subject control s robotic arm by fixate at stimulation ligh using SSVEP

### 2.2.3 Visual stimuli for the P300 brain-computer interface: A comparison of white/grey and green/blue flicker matrices[3]

P300 speller has mainly used white/grey flicker matrices as visual stimuli but they are not reducing the fatigue condition of the user. Parra and colleagues evaluated what colour is reducing the fatigue condition of the user. In their study, five single-color stimuli have been implemented. There are white, blue, red, yellow and green.

In this experiment, the green/blue chromatic flicker emerged as the safest and evoked the lowest rate of EEG spikes. The result showed that the accuracy rate was higher in response to the luminance chromatic flicker condition(LC) than in response to the luminance(L) or chromatic(C) flicker condition.

In conclusion, most users preferred green and blue make them feel less strain on the eyes. The subjects found that they could use green stimulus for longer periods as compared with red and blue in all frequency range.

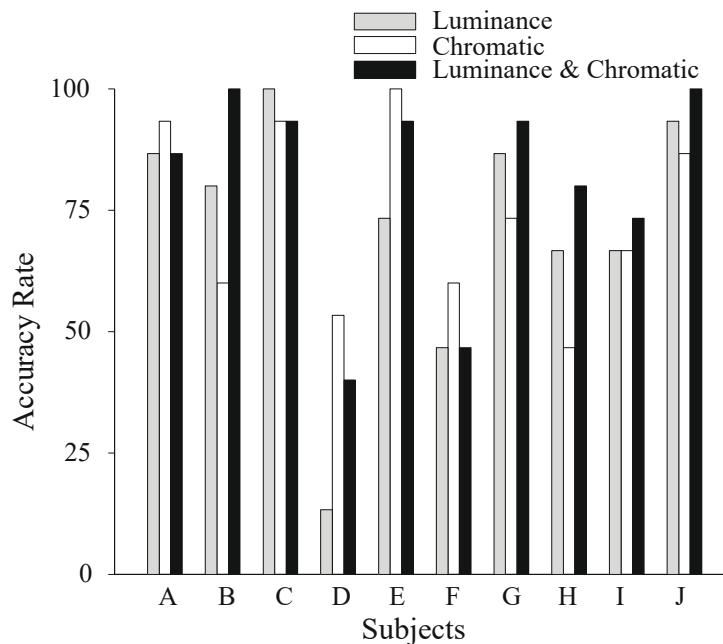


Figure 2.3: Accuracy rates for each subject. The accuracy rates in L, C, and LC conditions for each subject (AJ) are indicated by grey bars, white bars, and black bars, respectively.

### 2.2.4 Validation of the Emotiv EPOC EEG gaming system for measuring research quality auditory ERPs[4]

There are two kinds of equipment for obtaining a brain signal as follow: Researching headset and Gaming headset. Studying this research can be concluded that the Gaming headset can be used as well as the Research headset because of the result of an experiment showed that the characteristics of the Gaming headset's graph are very similar to the graph of Research headset.

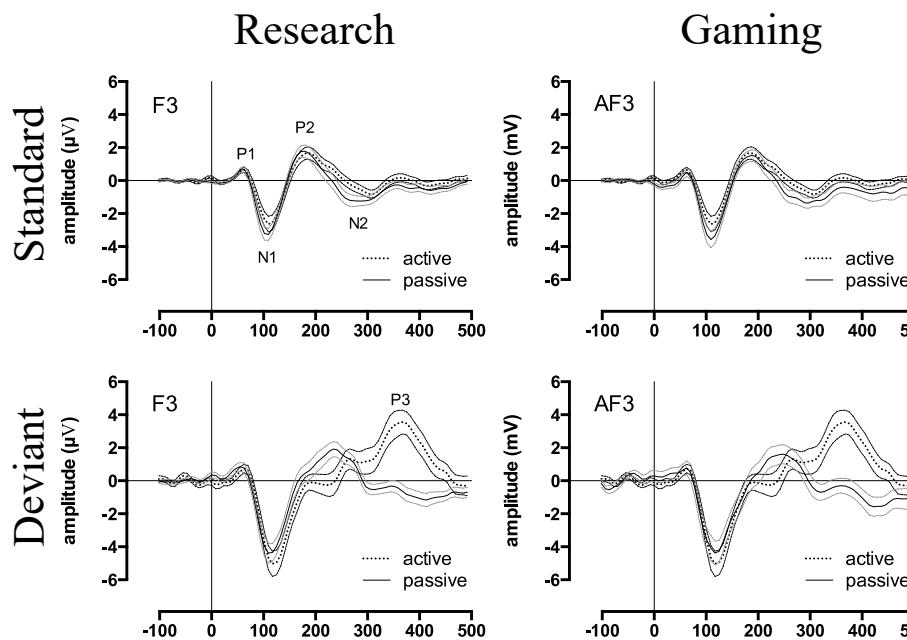


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### 2.2.5 c-VEP Brain-Computer Interface (BCI)[5]

Code-modulated visual evoked potentials (c-VEP) is one of the kinds of BCI. c-VEP uses pseudorandom code to modulate different visual stimuli. The figure 2.5 A shows the configuration and modulation of the c-VEP BCI system

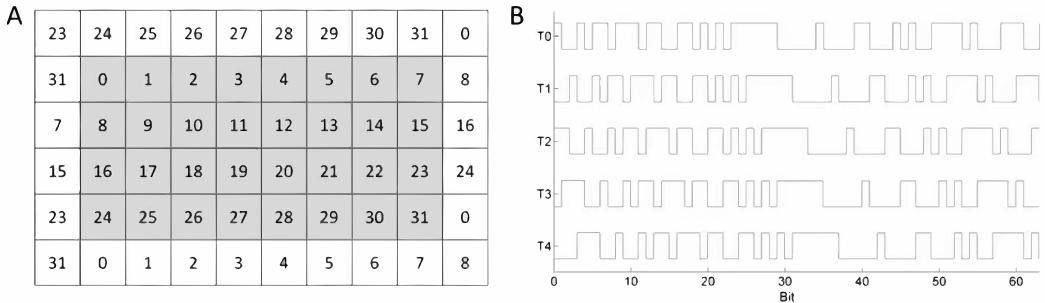


Figure 2.5: A: The gray area shows the 32 target stimuli with the number referring to number of target. The stimuli in white area are the complementary, which are synchronized to the target with same number. B: Modulation sequence for the first 5 targets. The sequence of a target is shifted by 2 bit in respect to its preceding target.

### 2.2.6 SSVEP-based brain computer interface using the Emotiv EPOC[6]

After the data was obtained from the headset, the two occipital channels were averaged together which help eliminated some of the noise between two channels. Because SSVEP is base on a particular frequency being present in the EEG data, the signal processing should be accomplished by using a Fourier transform to convert from the time domain into the frequency domain. After averaged the data. The result from converting is a frequency spectrum. Then, squaring the amplitude of each frequency component in this spectrum. This result was collected as active signal spectrum After that using a baseline removal method to further improve the results of signal isolation for the identification of SSVEP response. A baseline was recorded while the user was viewing a solid 50 percent gray screen without present any visual stimulation. After the data was recorded, Begin the method by using a Fourier transform with baseline spectrum and then reduce the noise by using a smoothing filter. Finally, the smoothed baseline spectrum was subtracted from the active signal spectrum. Finally, the result from this is the spectrum that can classify the observed data.

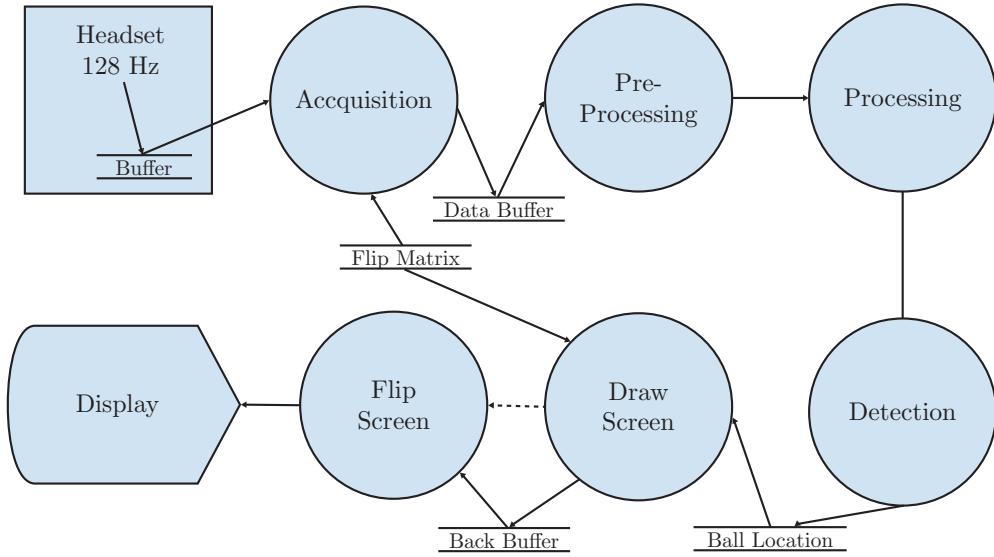


Figure 2.6: Diagram of data processing in EEG headset device [6]

### 2.2.7 A Survey of Stimulation Methods Used in SSVEP-Based BCIs[7]

The stimulus frequency can be divided into three bands. that is low (1-12 Hz), medium (12-30 Hz) and high (30-60 Hz). The largest SSVEP amplitudes were observed near 10 Hz followed by 16-18 Hz. Therefore, the most of the SSVEP-based BCIs use low and medium frequency bands. However, These two frequency bands have the disadvantage. First, the subjective evaluation shows that frequency between 5 and 25 Hz are more annoying than higher ones which visual fatigue would easily occur. Second, flash and pattern reversal stimuli can provoke epileptic seizures especially in the 15-25 Hz range. Third, the low-frequency band covers the alpha band (8-13 Hz) which can cause a considerable amount of false positives. All of these disadvantages can be avoided by using the higher frequency band.

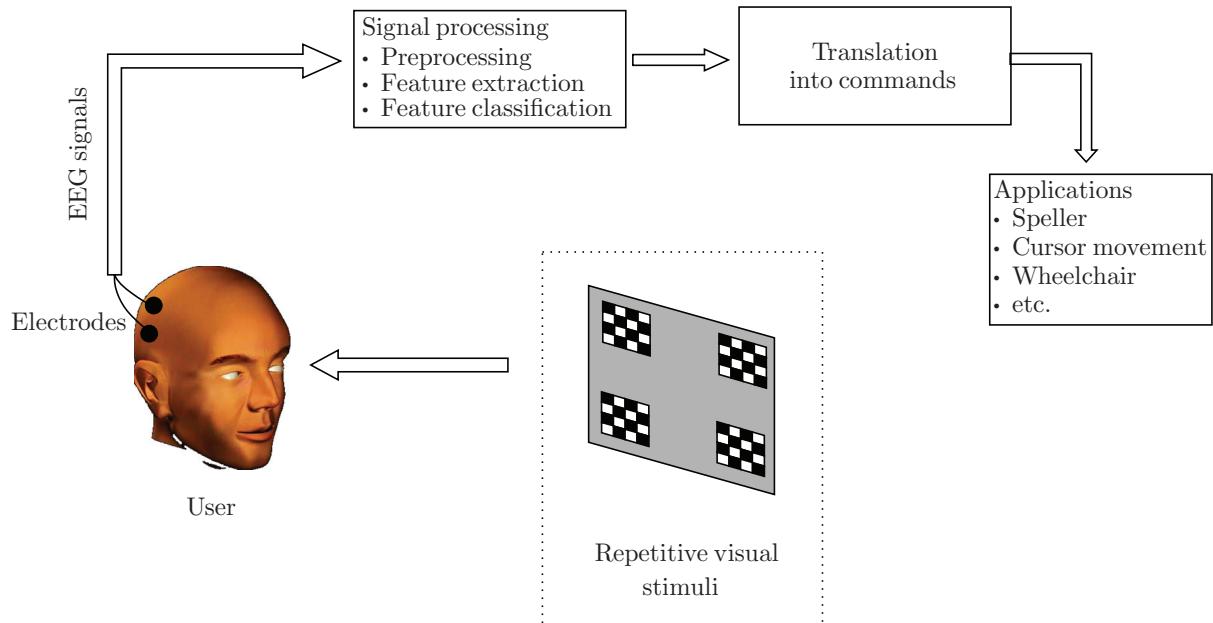


Figure 2.7: Functional model of an SSVEP-based BCI

### 2.2.8 Human EEG responses to 1-100 Hz flicker[8]

Human EEG responses to 1-100 Hz flicker is stimulated in visual cortex. Herrmann reported that the SSVEP responses exhibited resonance phenomena around 10,20,40 and 80 Hz

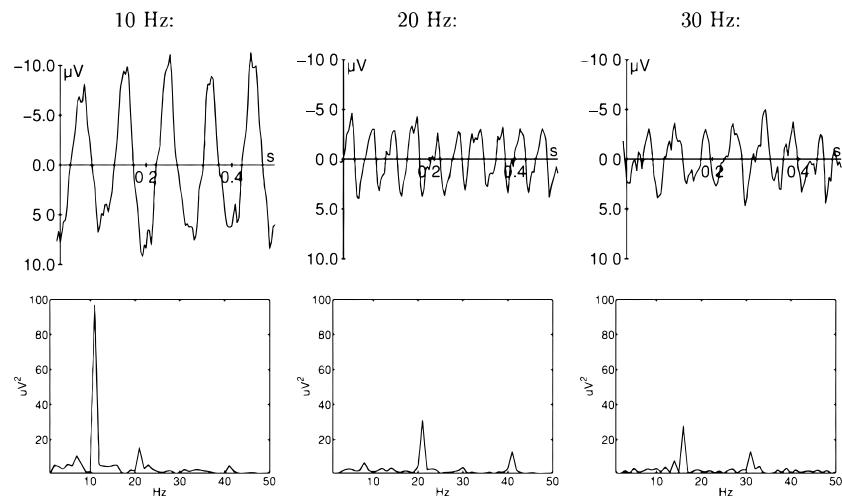


Figure 2.8: SSVEPs of a single subject in response to 10 Hz (left), 20 Hz (middle) and 30 Hz (right) stimulation (top row) and the corresponding FFT frequency spectra (bottom row)

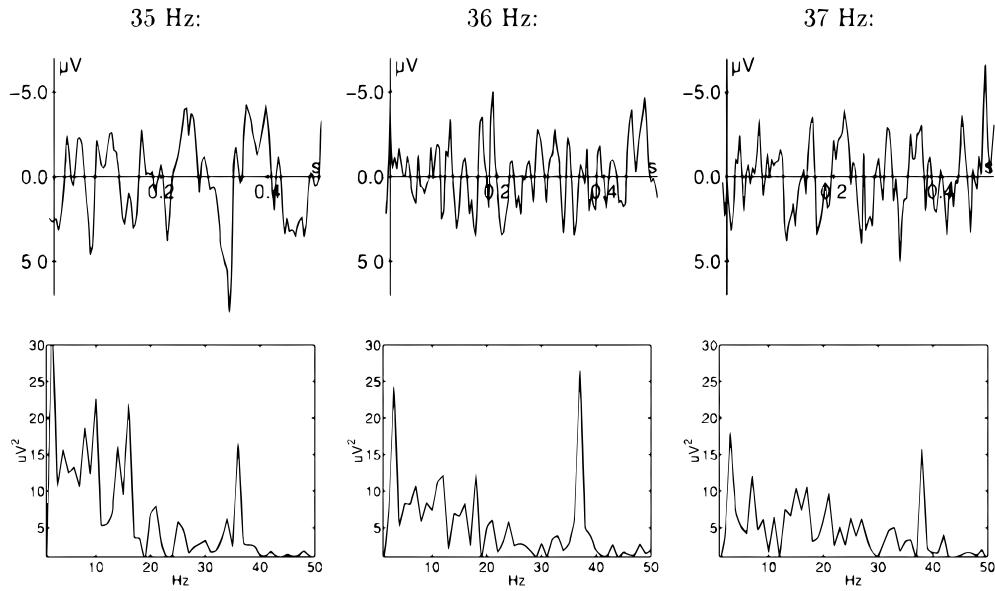


Figure 2.9: SSVEPs of a single subject in response to 35 Hz (left), 36 Hz (middle) and 37 Hz (right) stimulation (top row). The corresponding FFT frequency spectra show an increase of power at 36 Hz for 36 Hz stimulation (middle) as compared to adjacent frequencies (left and right)

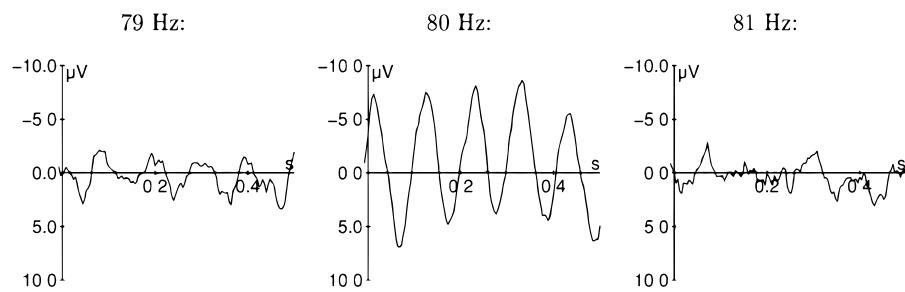


Figure 2.10: SSVEPs in response to flicker frequencies 79 (left), 80 Hz (middle) and 81 Hz (right). The VEP shows clear 10-Hz oscillations at 80 Hz which are not as prominent for the adjacent frequencies

# Chapter 3

## Background knowledge

### 3.1 Background knowledge

#### 3.1.1 Electroencephalography (EEG)

EEG is a technique for studying the electrical activities within the brain using electrodes that attach to the surface of the scalp. Wires attach these electrodes to a machine, which records the electrical impulses. The results are either printed out or displayed on a computer screen. A different pattern of electrical impulses can denote the various form of epilepsy[9]. It is also used to diagnose sleep disorders, coma, encephalopathies, and brain death.

Derivatives of the EEG technique include evoked potentials (EP), which involves averaging the EEG activity time-locked to the presentation of a stimulus of some sort (visual, somatosensory, or auditory). Event-related potentials (ERPs) refers to averaged EEG responses that are time-locked to more complex processing of stimuli. While Steady state visually evoked potential (SSVEP) refers to stable EEG waveform which has the same or multiple of the frequency of the visual stimulus



Figure 3.1: EEG signal

### 3.1.2 Event-related potential (ERPs)

Event-related potentials (ERPs) are very small voltages generated in the brain structures in response to specific events or stimuli. They measure brain response that is directed result of a specific sensory, cognitive or motor event. To see the brain's response to a stimulus, the experimenter must conduct many trials (usually in the order of 100 or more) and average the results together, causing random brain activity to be averaged out and the relevant waveform to remain, these stimuli can be visual, auditory, tactile and even olfactory and gustatory.

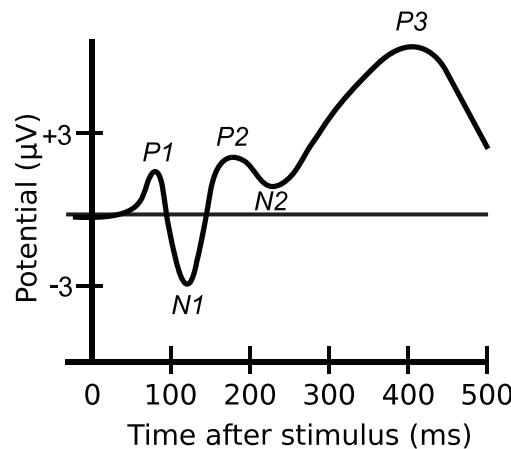


Figure 3.2: Averaged ERP waveform

### 3.1.3 Steady state visually evoked potential (SSVEP)

Steady State Visually Evoked Potentials or SSVEPs are EEG signals that are natural responses to visual stimulation. When the subject is in a meditative state, the brain will generate a stable EEG waveform. But when the subject's retina is stimulated by a visual stimulus ranging from 3.5 Hz to 75 Hz, the brain generates electrical activity at the same or multiple of the frequency of the visual stimulus.

### 3.1.4 Cerebral Cortex

The human brain consists of several parts. The part that makes humans more intelligent than any animals is the cerebrum. The cerebrum is the largest part of the human brain. The physician divides the cerebrum into four parts. There is frontal lobe, parietal lobe, temporal lobe and occipital lobe. The frontal lobe is located at the front of the brain. It controls the creative thought, thinking, problem solving and muscle movements. The parietal lobe is located behind the frontal lobe. It is involved in the visual functions, languages, reading and sensory controls. The temporal lobe is situated beneath both frontal lobe and parietal lobe. It is implicated in control of memory, speech and hearing languages. The occipital lobe is located at the back of the brain. It is related with human vision control. Due to our project study about the human visualization, so we focus on the occipital lobe.

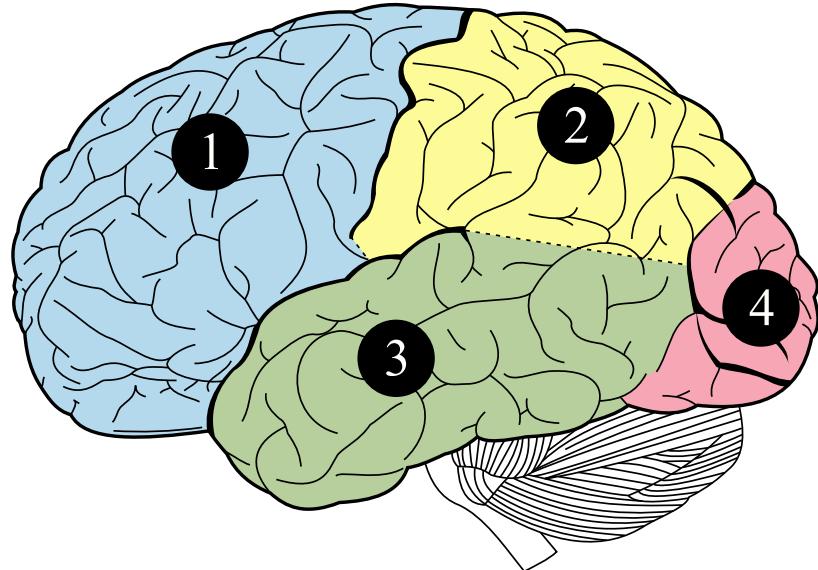


Figure 3.3: 1.Frontal lobe, 2.Parietal lobe, 3.Temporal lobe, 4.Occipital lobe

### 3.1.5 Feature scaling (normalization)

Feature scaling is used to normalize the range of independent variables or features of data to have the properties of a standard normal distribution. to make in some machine learning algorithms work properly. it is usually performed while in the data pre-processing step. The calculation is determined the distribution mean and standard deviation for each feature. Next, we subtract the mean from each feature. Then divide the values which already subtracted by mean, by its standard deviation.[10]

$$x' = \frac{x - \bar{x}}{\sigma}$$

### 3.1.6 Uniform distribution[11]

The Uniform or Rectangular distribution has random variable restricted to a finite interval and has constant density over the interval.

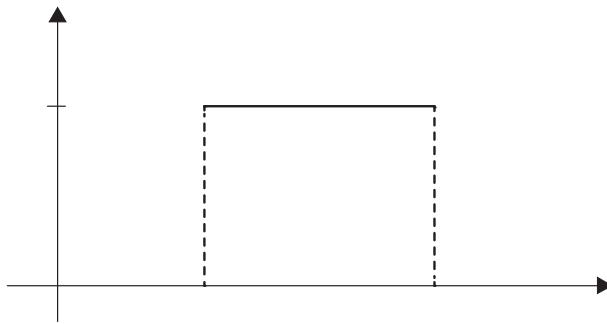


Figure 3.4: An illustrate of uniform distribution

The function  $f(x)$  is defined by:

$$f(n) = \begin{cases} \frac{1}{b-a} & \text{if } a \leq x \leq b \\ 0 & \text{otherwise} \end{cases}$$

### 3.1.7 10/20 international system

The 10/20 international system is a recognized method to describe and apply the location of scalp electrodes in the context of EEG test. This system is based on the relationship between the location of an electrode and the underlying area of cerebral cortex. The 10/20 international system refers to the actual distance between adjacent electrodes are either 10 % or 20 % of the total front-back or right-left distance of the skull. Each site has letters to identify the lobe and a number to identify the hemisphere location. The letters F stands for frontal lobe, T is temporal lobe, C is central lobe, P is parietal lobe, and the last one is O means occipital lobe. Even numbers refer to electrode positions on the right hemisphere and odd numbers refer to the electrode positions on the left hemisphere.

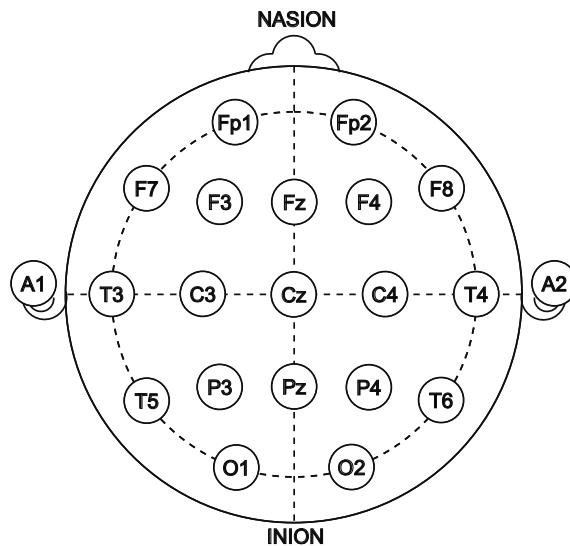


Figure 3.5: International 10-20 locations system

### 3.1.8 Standard deviation

Standard deviation(SD) is measuring a distribution of the collection of data.if the SD is high, the distribution is high. if the SD is low, the distribution is low.

### 3.1.9 Low-pass filter

The Low-pass filer only allow the frequency signal that is below cut-off frequency

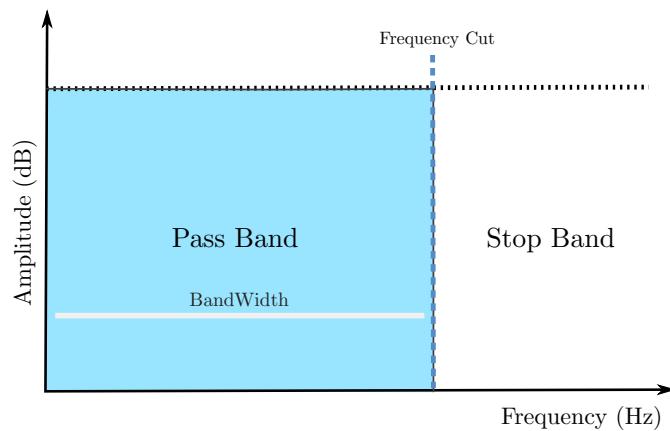


Figure 3.6: Low-pass filter

### 3.1.10 High-pass filter

The High-pass filter only allow the frequency signal that is higher than cut-off frequency

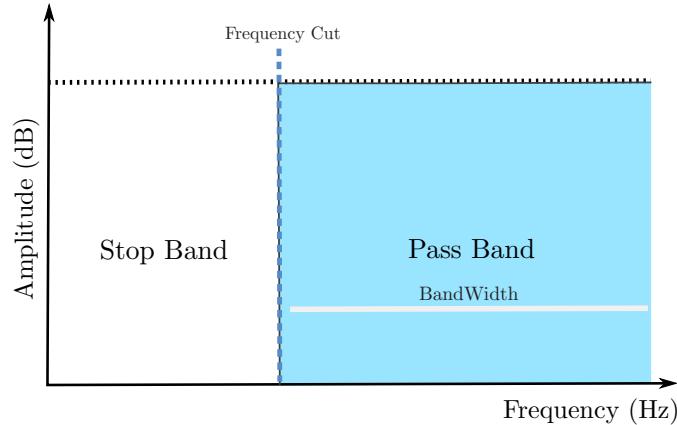


Figure 3.7: High-pass filter

### 3.1.11 Band-pass filter

The Band-pass filter only allow the frequency signal that is in the specified range

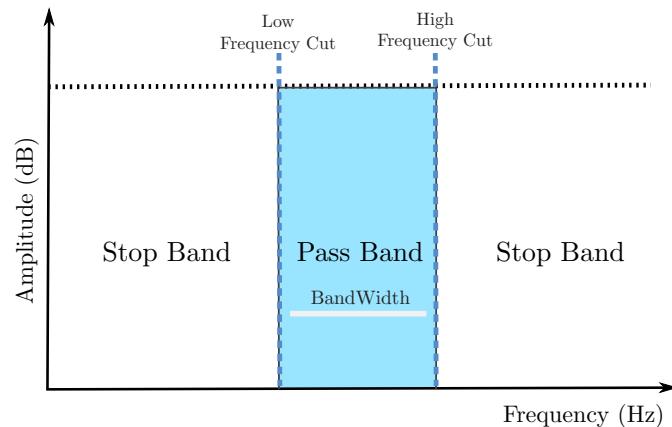


Figure 3.8: Band-pass filter

### 3.1.12 EEG band

EEG band determined as a fixed range of wave frequency and amplitudes over a time scale. In Scientific field classify into 5 types. that are,

1.Delta band is the lowest frequency in the range of 0-3 Hz. It occurs during deep sleep

2.Theta band is in the frequency range between 4-7 Hz. We can investigate the situation of intuitive,creative and imagery.

3.Alpha band is in the frequency range between 8-12 Hz. It occurs during the situation that our mind calm,so we can obtain the strongest EEG brain signal in this frequency range.

4.Beta band is in the frequency range between 12-30 Hz. We can investigate this band at the situation of memory and problem-solving.

5.Gamma band is in the frequency range between 30-100 Hz. This wave may be implicated in creating the unity of conscious perception.

### 3.1.13 Fourier transform

The Fourier transform is the mathematical model which uses to transform from the time signal into the frequency domain.

### 3.1.14 Short-time Fourier transform

The short-time Fourier transform(STFT) divide a long time signal into shorter groups before computing the Fourier transform on each group.

## 3.2 Material background

### 3.2.1 Emotiv EPOC headset[12]

The wireless headset Emotiv EPOC research edition, it records EEG data in 14 channel of International 10-20 Locations system with Sequential sampling rate at 128 per second (2048 Hz internal) with resolution 14 bit (16 bit Analog to digital converter, 2-bit instrumental noise floor discarded), the bandwidth is in range 0.2 to 45 Hz with digital notch filters at 50Hz and 60 Hz



Figure 3.9: Emotive Headset

### 3.2.2 Gravitech Gerora LED[13]

The full-color Light-emitting diode (LED) driver which is used for stimulating the subject has an LED light WS2812 model mounted on. It can chain connected with the same model. It can be controlled by a programmed Arduino. The luminous intensity is 550 to 700 mega candela for red, 1100-1400 mcd for green, and 200-400 mcd for blue.

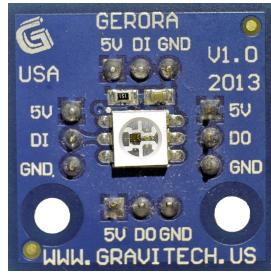


Figure 3.10: Gravitech Gerora LED base on WS2812

### 3.2.3 Arduino UNO[14]

The Arduino Uno is a microcontroller board based on the ATmega328 that is an open source platform. It has 14 input/output pins, 6 analog inputs, a 16 MHz quartz crystal, a USB connection, a power jack, an ICSP header and a reset button. The USB connection for upload the software into the Arduino and VCC or supply for connecting the peripheral circuit.

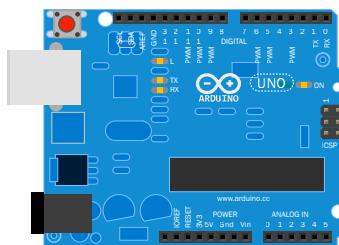


Figure 3.11: Illustrate of Arduino UNO board

# **Chapter 4**

## **Requirement analysis and Software design**

### **4.1 Requirement for application**

#### **4.1.1 Functional requirement**

- The system allows user to control machine using brainwave.
- The system allows user to check the EEG headset status.
- The system allows administrator to see the authentication result.
- The system allows administrator to see real-time graph of each band.
- The system allows administrator to check the EEG headset status.
- The system allows user to save information into the system.
- The system shows the analysis result.

#### **4.1.2 Non-Functional requirement**

- The system uses C# language to develop software.

- The system uses Windows Presentation form and Visual studio 2015 to create user interface.
- User friendly: The UI is look clear, simple, easy to use and understand.
- Performance: The system shows and records the EEG brainwave in real-time.

## 4.2 Use case Diagram

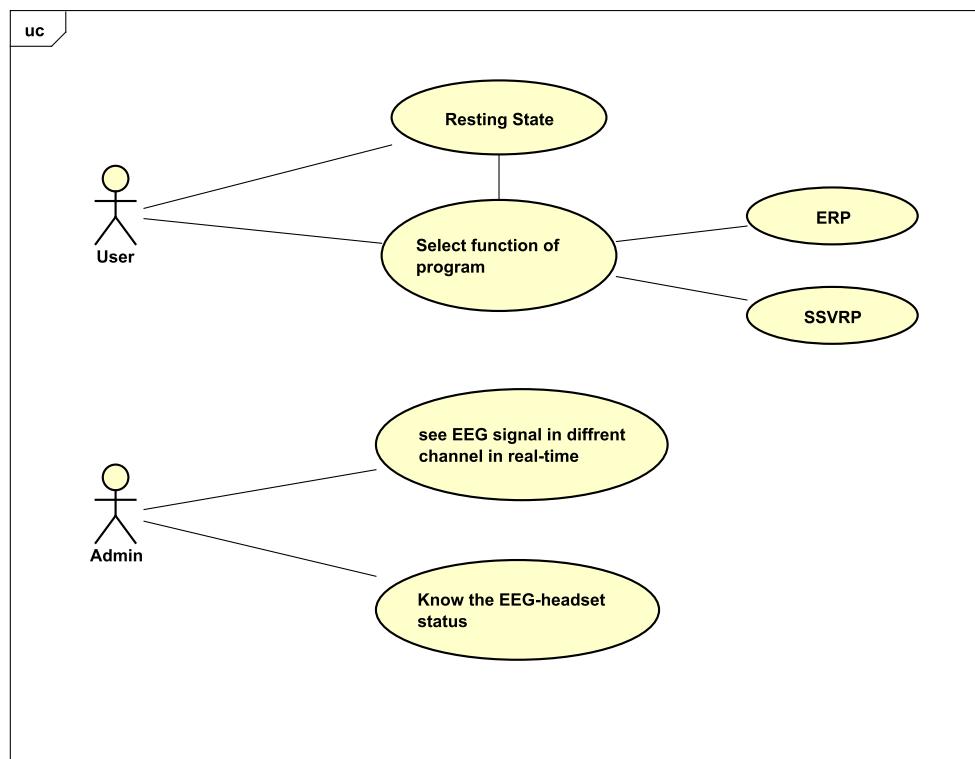


Figure 4.1: Use case diagram

## 4.3 Activity Diagram

### 4.3.1 Activity diagram of resting state

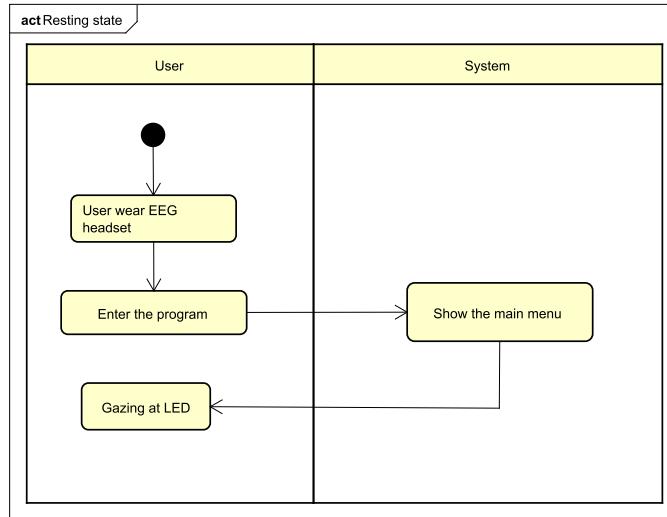


Figure 4.2: Activity diagram of resting state

### 4.3.2 Activity diagram of ERP

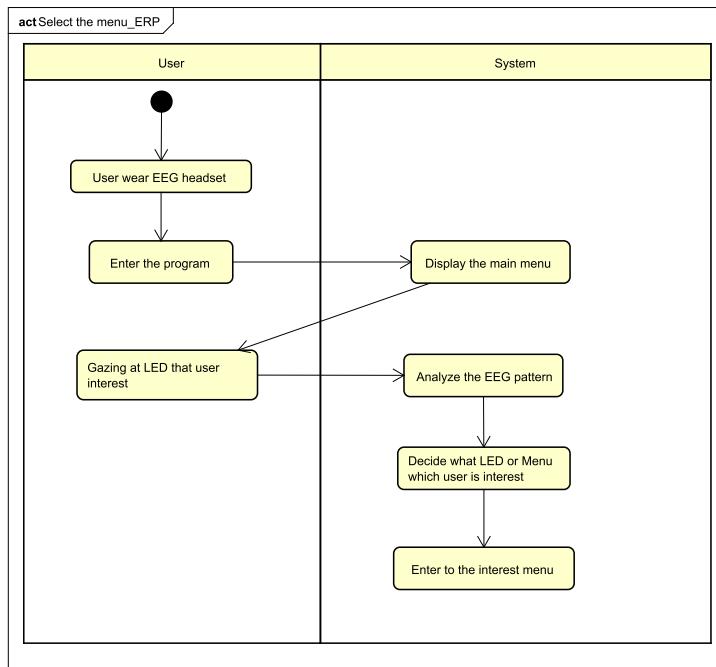


Figure 4.3: Activity diagram of ERP

### 4.3.3 Activity diagram of SSVEP

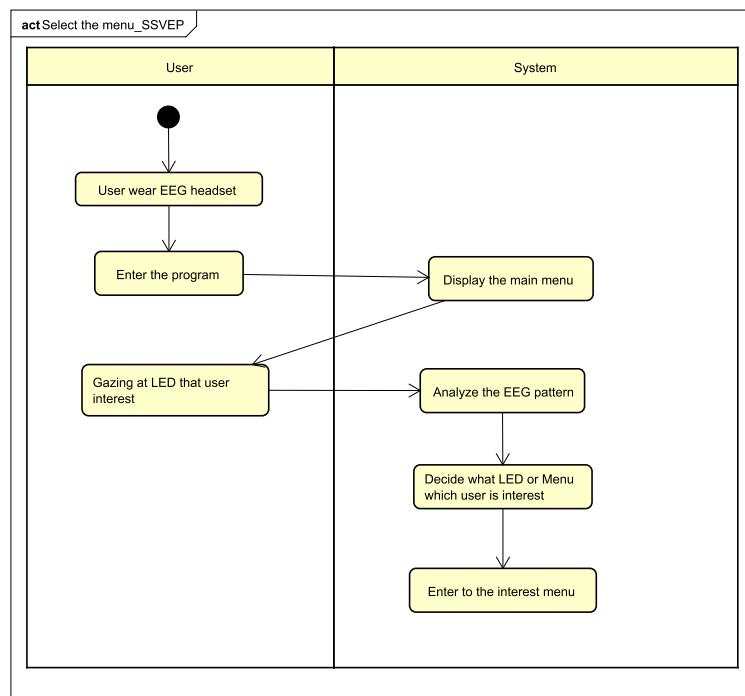


Figure 4.4: Activity diagram of SSVEP

# Chapter 5

## Software design

### 5.1 System architecture

#### 5.1.1 System architecture

In our project, we use two algorithms to stimulate the subject. First use ERPs which is different in stimulating time. Second is SSVEP which difference in stimulating frequency.

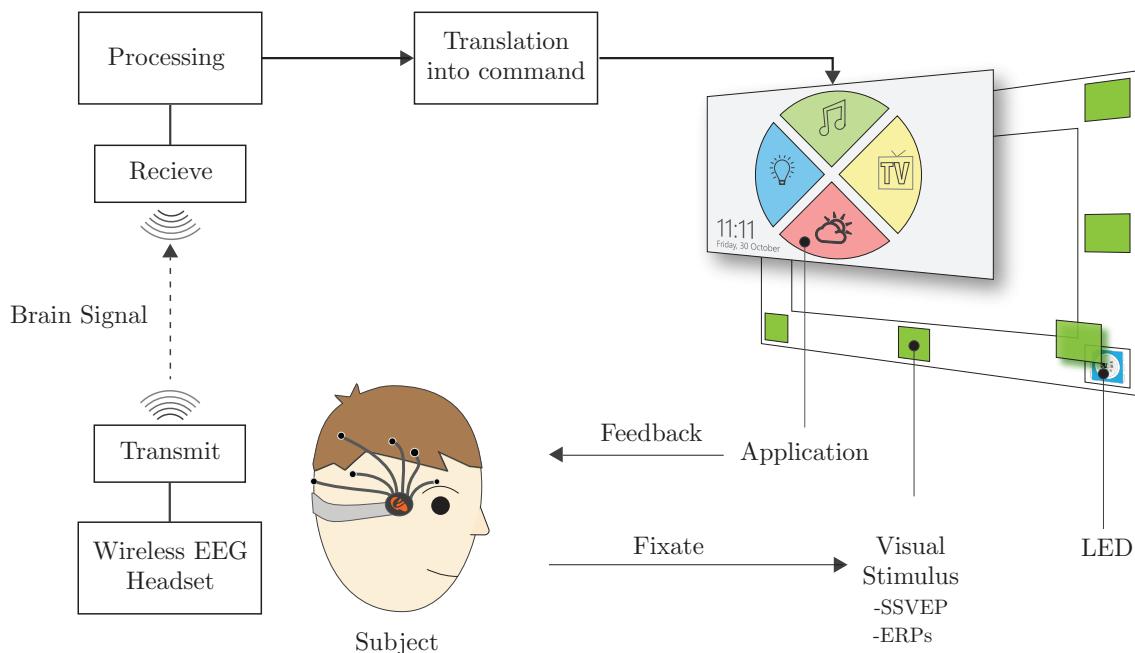


Figure 5.1: System architecture design

### 5.1.2 Class Diagram

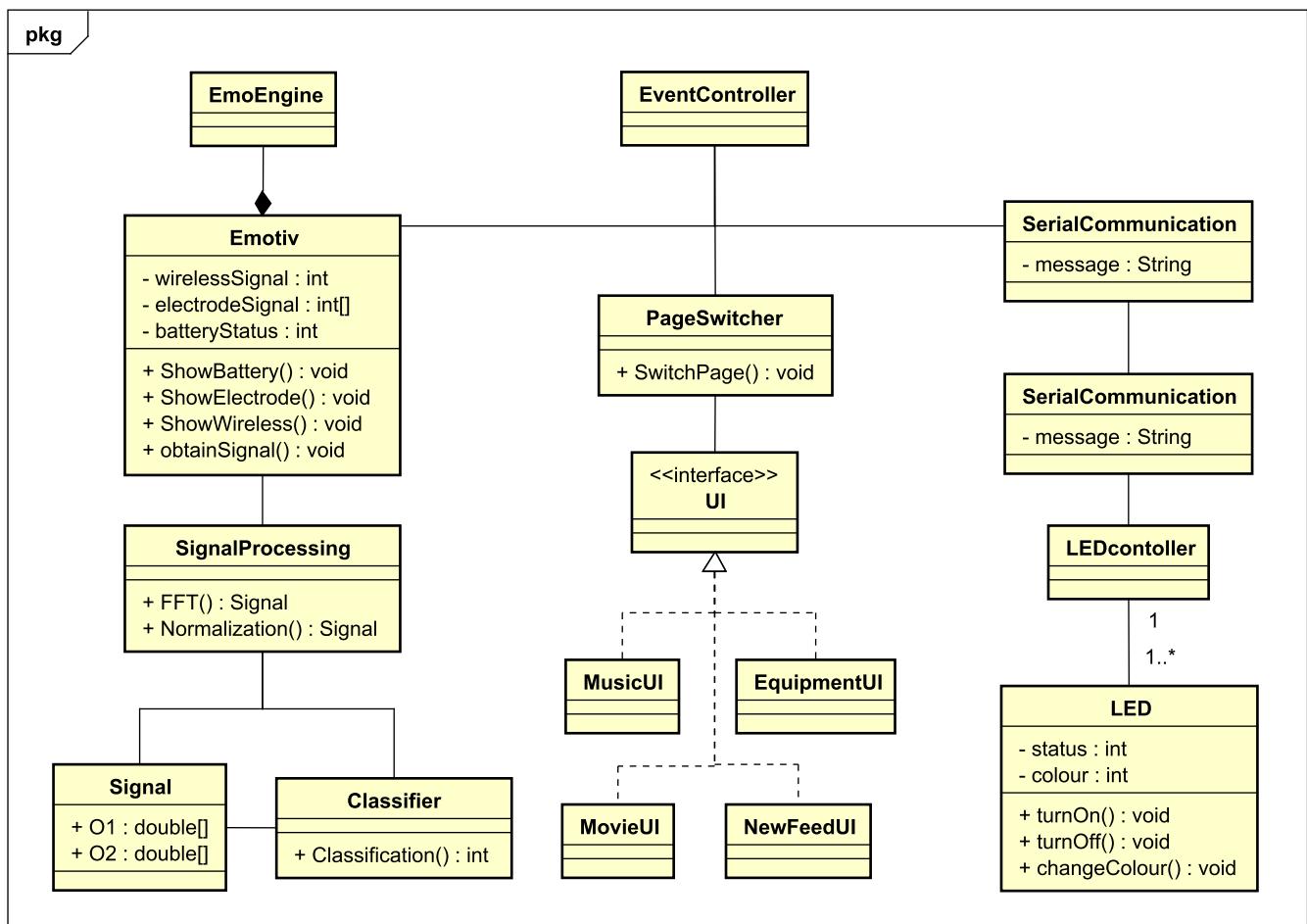


Figure 5.2: Class diagram

class diagram description, EmoEngin is a library use to connect to Emotiv SDK.

# **Chapter 6**

## **Development**

### **6.1 Chamber for the SSVEP experiment**

# **Chapter 7**

## **Experiments and results**

### **7.1 Hardware for the experiment**

# **Chapter 8**

## **Conclusion**

### **8.1 Summary of Thesis Achievements**

Summary.

### **8.2 Applications**

Applications.

### **8.3 Future Work**

Future Work.

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