MindGames

Project Breakdown

Brandon Knotek (bk11@hood.edu)
Walid Muhammad (wm5@hood.edu)
Jack Wilder (jmw38@hood.edu)

Our Goal

To explore the viability of neurofeedback as a means for controlling real-time interactive experiences

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To explore the viability of neurofeedback as a means for controlling real-time interactive experiences

In other words, play a game just by thinking commands; no mouse and keyboard required!

Why Bother?

• Accessibility - Physical disabilities may limit use of traditional input methods

 Communication - Certain conditions make movement and speech impossible (Locked In Syndrome)

So How Does It Work?

1. Brain activity is recorded and sent to a computer using special hardware

2. The computer processes the activity and translates it to commands

3. The commands are sent to the target software and executed

Unraveling the Brain

All brain activity is electrical

 Electroencephalography (EEG) uses electrodes placed on the scalp to measure this activity



Figure 1 - An example of EEG hardware

Unraveling the Brain

 Different thoughts, actions, and mental states produce unique electrical patterns

These patterns are different for any given person



Figure 2 - Sample EEG output

Building a Neural Fingerprint

 The computer must be trained to recognize an individual's unique patterns

 Using machine learning (ML), the computer is taught to associate labeled patterns with commands

```
Windows PowerShell
(i) Windows Terminal can be set as the default terminal application in your settings. Open Settings
Next trial: Neutral (13 / 50)
        Ready...
         Go
Next trial: Hard Blink (14 / 50)
        Readv...
        Go
Next trial: Left Arm (15 / 50)
Next trial: Left Arm (16 / 50)
        Readv...
Next trial: Jaw Clench (17 / 50)
        Ready...
Next trial: Left Arm (18 / 50)
        Readv...
         Go
Next trial: Neutral (19 / 50)
        Ready...
```

Figure 3 - Our calibration module collecting training data

Command Classification

 With enough examples, the ML model is able to classify subsequent patterns on its own

 Using probability, it determines the command most likely to correspond to a given pattern

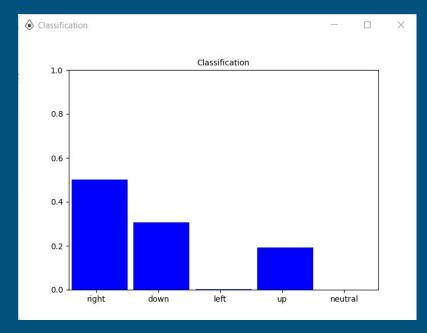


Figure 4 - A visualization of the ML model's classification output

Playing the Game

 The most likely command is sent to the game

 The command is executed and Pac Man moves in the desired direction

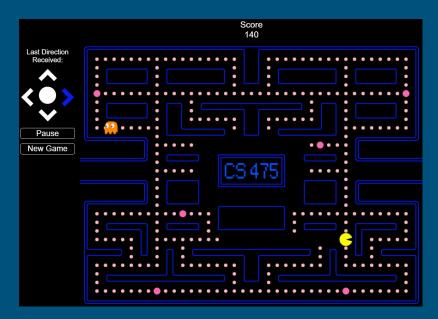


Figure 5 - Pac Man in action

System Overview

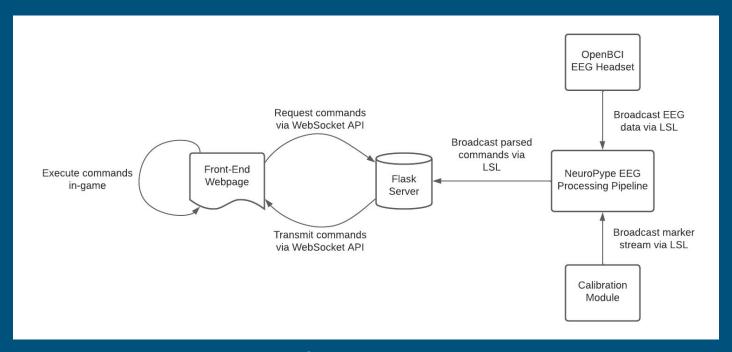


Figure 6 - A diagram of the MindGames system architecture

System Overview - Communication

- Lab Streaming Layer (LSL) Synchronization and transmission of real-time continuous measurements
 - EEG data and training labels

- WebSocket Continuous communication between a webpage and a server
 - Game commands

System Overview - Processing Pipeline

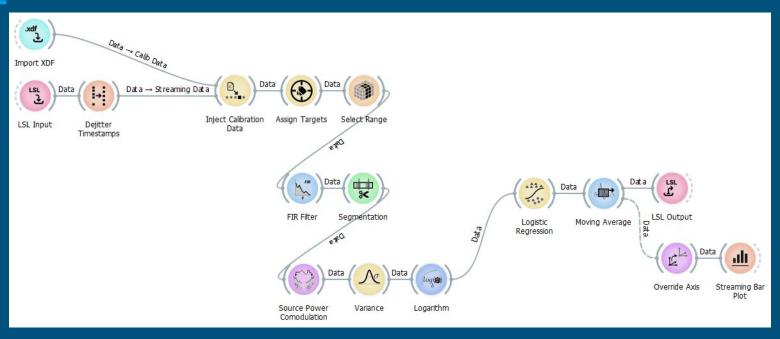


Figure 7 - The MindGames EEG processing pipeline

Processing Pipeline - Data Acquisition

 Get EEG data and ML training labels into the pipeline

 Assign training labels to commands

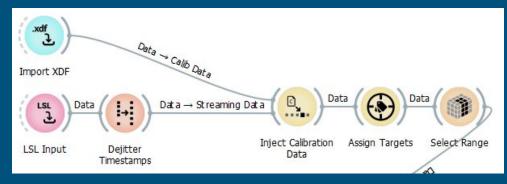


Figure 8 - EEG pipeline data acquisition component

Processing Pipeline - Preprocessing

Filter out electrical noise

 Designate length of data "chunks" to associate with labels

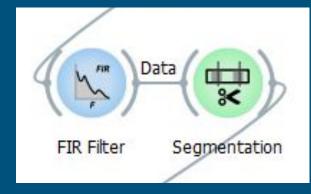


Figure 9 - EEG pipeline preprocessing component

Processing Pipeline - Feature Extraction

Extract EEG signal components that exhibit the most variance between commands

 Reduce the volume of data to make ML classification more efficient



Figure 10 - EEG pipeline feature extraction component

Processing Pipeline - Classification

 Use extracted features to determine command probabilities

 Maintain a moving average of probabilities to reduce noise

Output probabilities for further use

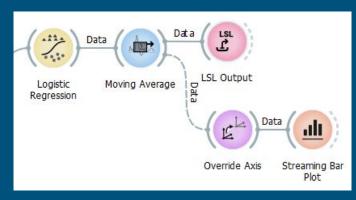


Figure 11 - EEG pipeline classification component

Challenges - Hardware

 EEG data collected using OpenBCI Ultracortex headset

 Difficult to secure, small movements cause it to shift position

 Electrodes move with the headset, generating spikes and invalidating training data



Figure 12 - Ultracortex headset, profile view

Challenges - Hardware

 Limited to 8 electrodes (professional setups may use up to 128)

 Electrodes are not secured to the scalp

RESULT: Low quality data (and not enough of it)

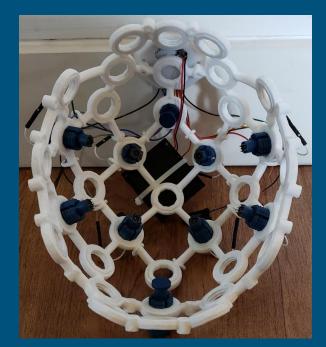


Figure 13 - Ultracortex headset, bottom view

Challenges - Software

 ML training takes 8 minutes, results become useless as electrodes shift with headset

 Noise and irrelevant spikes (e.g., blinking, swallowing) must be filtered out without affecting relevant signals

Everything must happen with minimal latency for the game to feel responsive

Next Steps

• Fine-tune our noise reduction and ML classification techniques

Devise a system to secure the headset in place

 Transition from classifying movements to classifying imagined movements (motor imagery)

Questions?