

Final Year Project 2024-2025

EEG OPERATED WHEEL CHAIR



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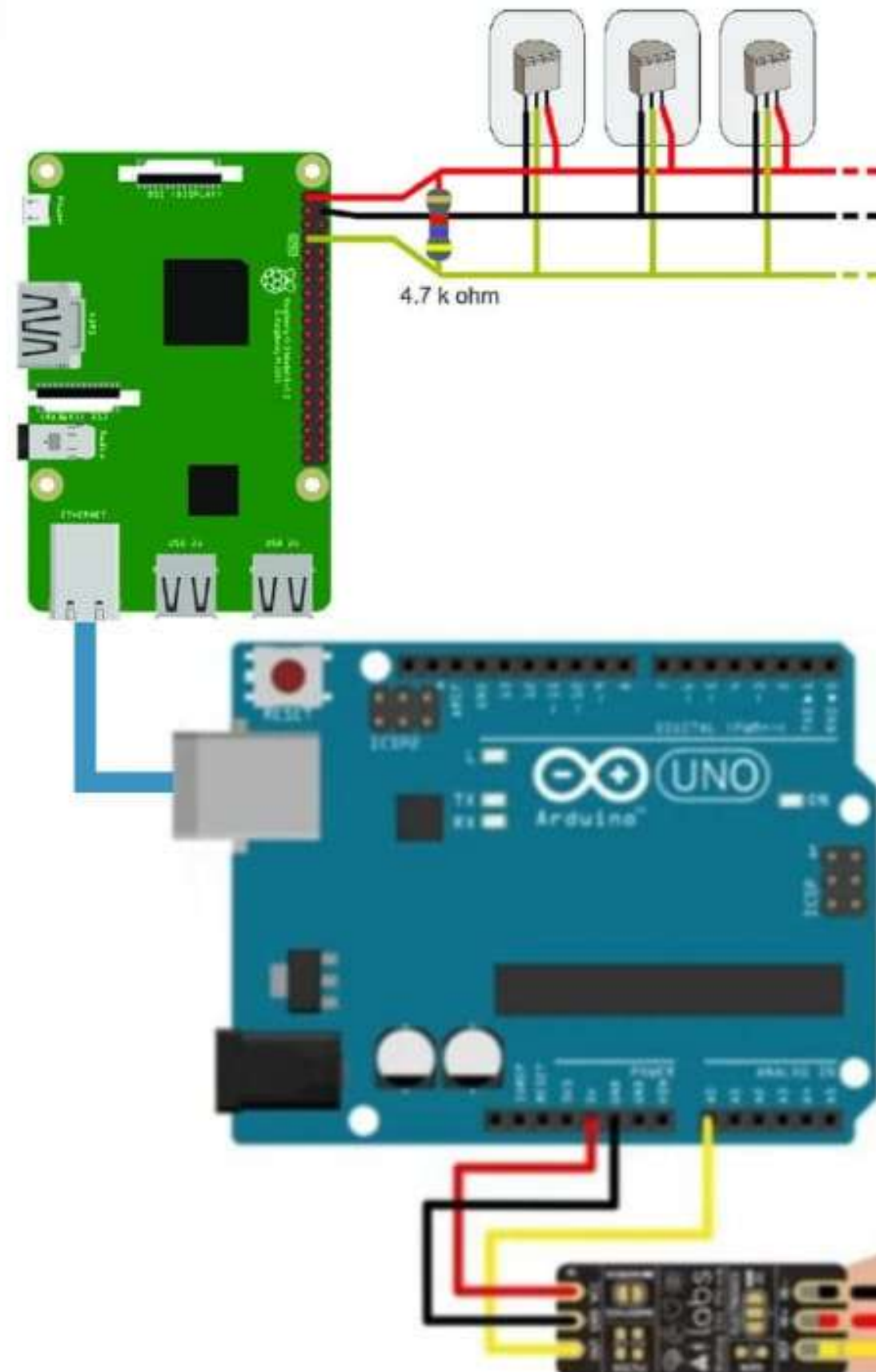
Prof. Tushar Mohije

Why Choose the BioAmp EXG Pill for EEG-Operated Wheelchair?

- Versatile & Precise: *Records high-quality biopotential signals (EEG, ECG, EOG, EMG).*
- Highly Compatible: *Works with various MCUs and SBCs (e.g., Arduino, ESP32, Raspberry Pi).*
- High Input Impedance: *$10^{12}\Omega$ ensures clean, accurate signals.*
- Configurable: *Supports different biopotentials and electrode setups (default: EEG, 3 electrodes).*
- Open Source: *Customizable hardware and software.*
- Compact Size: *Small and lightweight (25.4×10.0 mm).*



CIRCUIT DIAGRAM



v 1.0b

FRONTAL
CORTEX
RECORDING

Collecting Data using Python script

```
import serial
import csv
import time
import datetime
```

```
COM_PORT = 'COM5' # Arduino's COM port
BAUD_RATE = 115200 # Must match the Arduino's BAUD_RATE
```

```
# Open the serial connection
ser = serial.Serial(COM_PORT, BAUD_RATE)
```

```
# Create a CSV file to save the data
with open('signal.csv', 'a', newline='') as csvfile:
    csvwriter = csv.writer(csvfile)
```

```
# Set the maximum duration of the data collection (in
seconds)
max_duration = 180
```

```
start_time = time.time()
```

```
print("Collecting data...")
```

```
while time.time() - start_time < max_duration:
    # Read a line of data from the Arduino (until a newline character)
    data = ser.readline().decode("latin-1").strip()

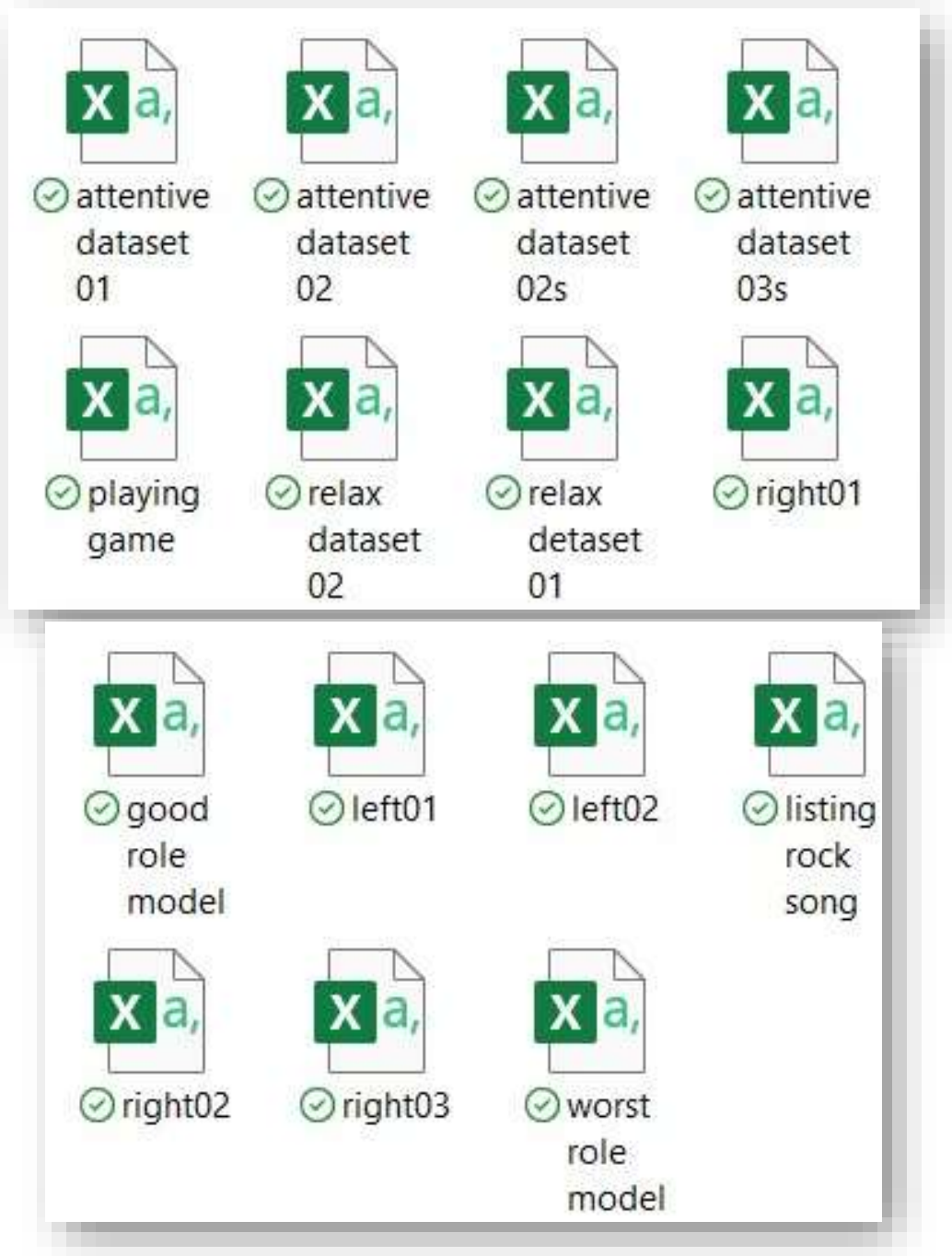
    # Get the current timestamp
    current_time = datetime.datetime.now().strftime('%Y-%m-%d %H:%M:%S.%f')

    # Split the data into a list of values using the comma as a delimiter
    values = data.split(',')

    if len(values) > 0 and values[0].isdigit():
        # Save the data to the CSV file along with the timestamp
        csvwriter.writerow([current_time, values[0]])

ser.close()
```

Datasets



Datasets Generation

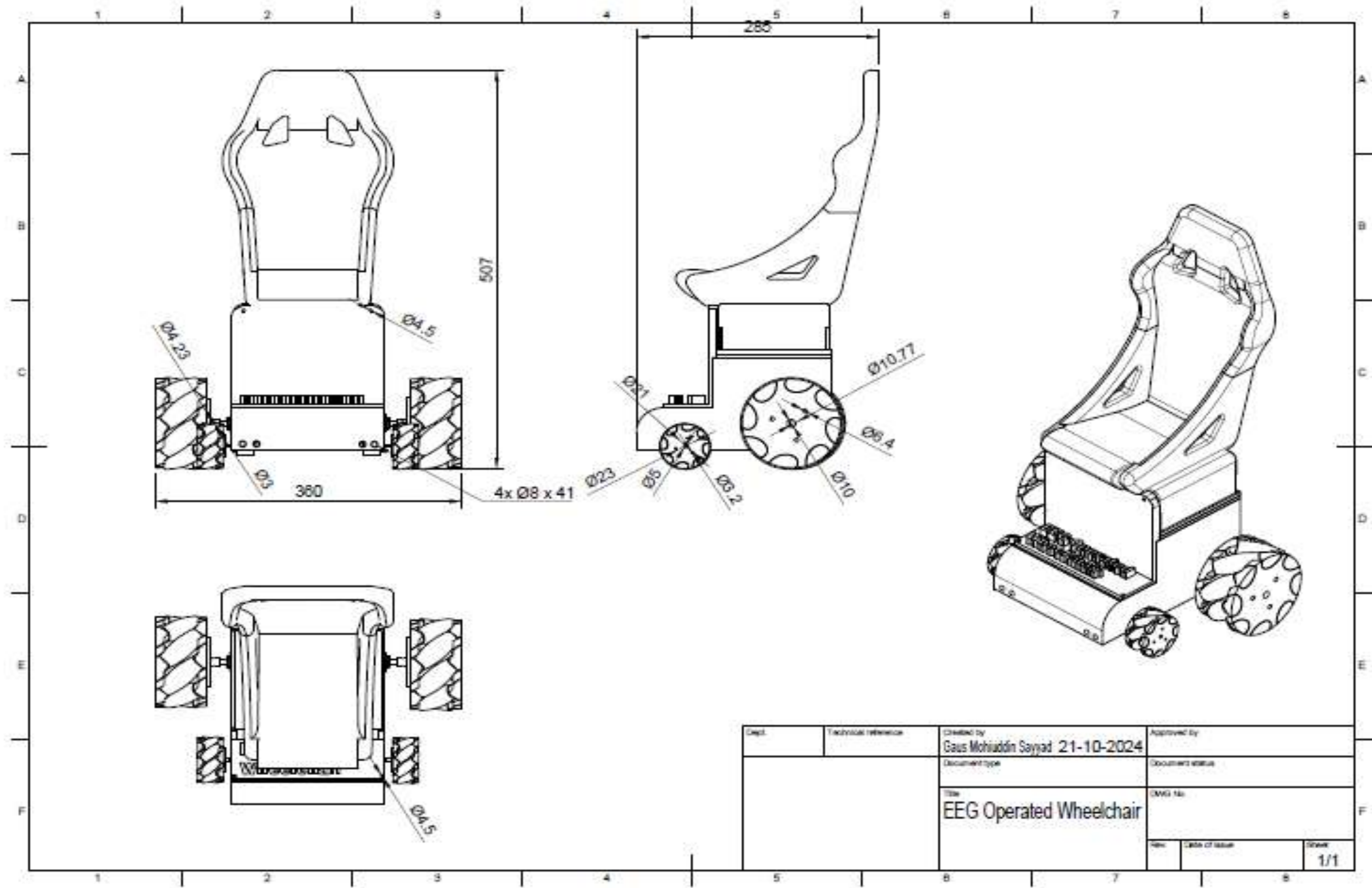
	26:50.1	519
0	26:50.1	507
1	26:50.1	502
2	26:50.1	483
3	26:50.1	506
4	26:50.1	492
...
159994	06:49.7	785
159995	06:49.7	678
159996	06:49.7	780
159997	06:49.7	796
159998	06:49.7	784
159999 rows × 2 columns		

Raw Data

	519	label
0	507	0.0
1	502	0.0
2	483	0.0
3	506	0.0
4	492	0.0
...
159994	785	3.0
159995	678	3.0
159996	780	3.0
159997	796	3.0
159998	784	3.0

Clean Labeled Data

CAD Designing



Brain Wave Data Collection

Directional Thinking:

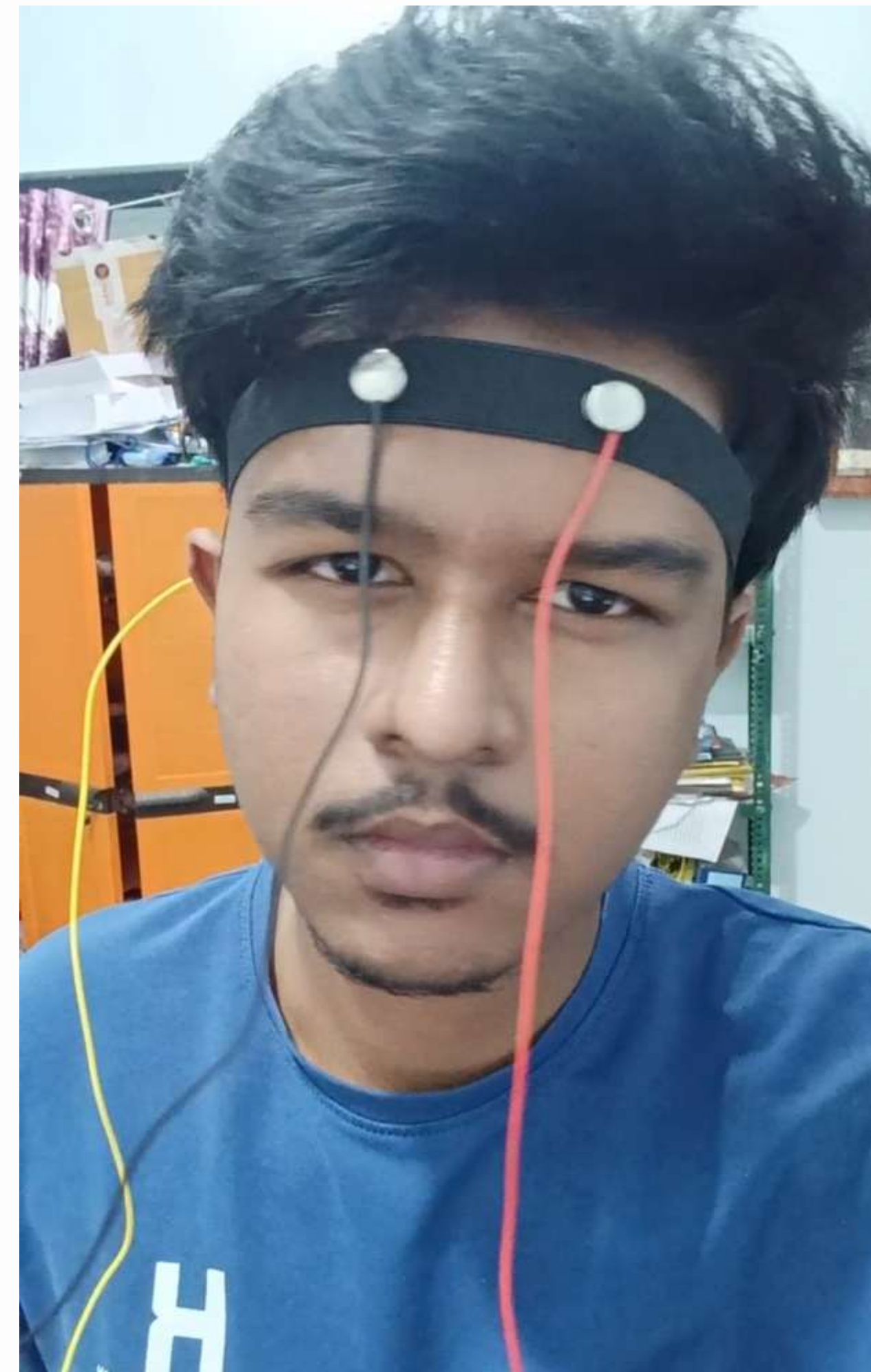
Subject think left, right, back, and front for 20 minutes each, repeated 3 times.

Activity-Based Data:

Brain waves recorded during activities like:

- *Playing games*
- *Listening to pop music*
- *Listening to devotional songs*
- *Thinking about an ideal character*
- *Thinking about a bad character (e.g., enemies)*

Data Collected: 1,18,98,816 instances



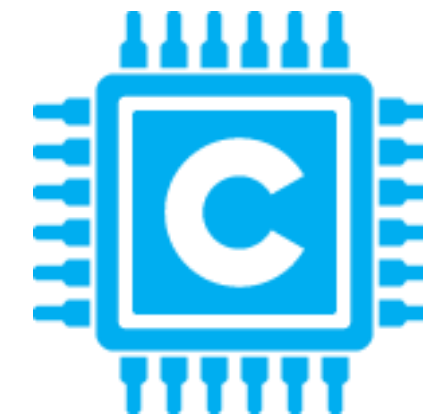
Data Processing & Model Training

Challenge: Large dataset of 1,18,98,816 instances.

Solution: Train a machine learning model to differentiate thinking patterns.

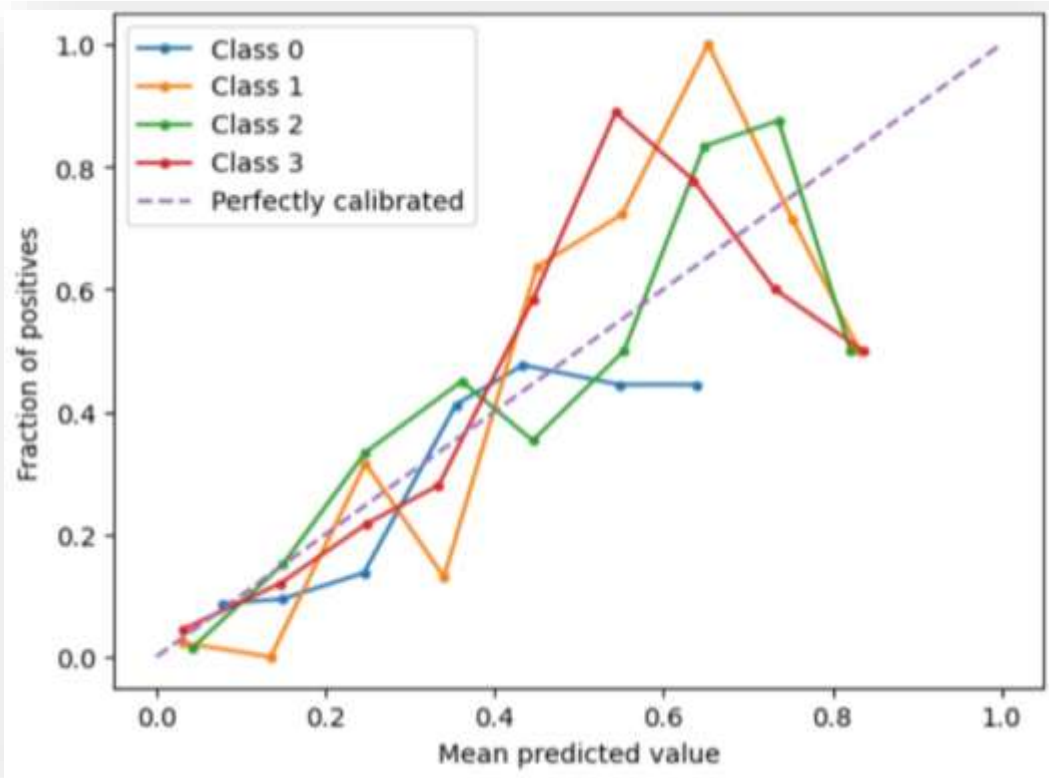
Tools Used:

- *NumPy: Numerical operations*
- *SciPy: Scientific computing*
- *Pandas: Data manipulation*
- *PySerial: Serial communication*
- *PyAutoGUI: GUI automation*
- *Scikit-learn: Machine learning algorithms*
- *Embedded C: Motor Controls*



Final Results

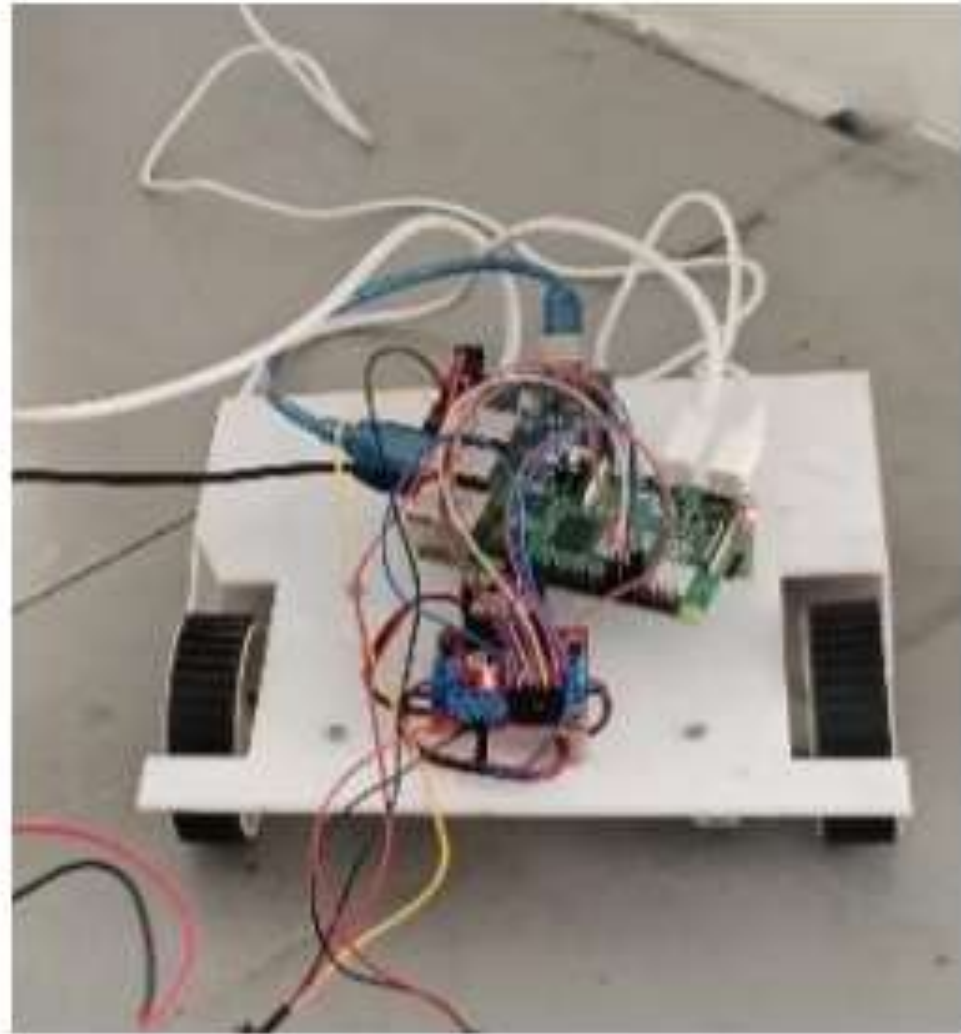
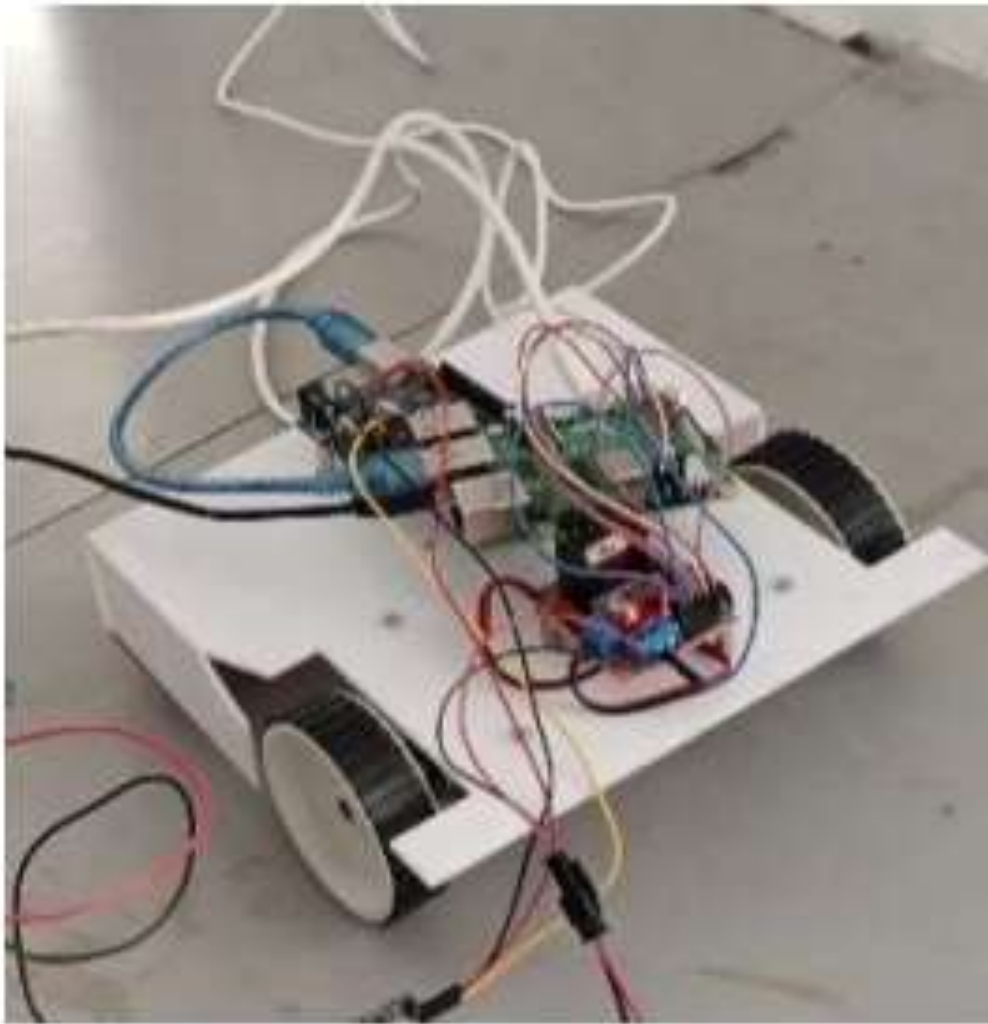
	E_alpha	E_beta	E_theta	E_delta	alpha_beta_ratio	peak_frequency	spectral_centroid	spectral_slope	label
count	623.000000	623.000000	623.000000	623.000000	623.000000	623.000000	623.000000	623.000000	623.000000
mean	8.826992	26.035070	3.360518	17.462140	0.391292	6.484926	12.494754	-10.750196	1.492777
std	8.789737	21.978750	3.966043	23.737435	0.298327	8.362149	3.453279	0.464050	1.118190
min	0.335261	3.880288	0.100366	0.120896	0.014671	0.000000	3.490535	-13.058867	0.000000
25%	3.356193	11.130909	1.046485	3.783032	0.196005	0.998051	10.001617	-11.006133	0.000000
50%	5.850757	17.813897	2.001234	9.007126	0.310454	0.998051	12.839840	-10.705366	1.000000
75%	10.649203	30.530237	3.965997	22.288495	0.485042	13.473684	15.308600	-10.418100	2.000000
max	63.977757	144.413950	40.652637	183.388871	2.511795	29.941520	19.627692	-9.870829	3.000000



```
{'E_alpha': 8.768172935724877,
'E_beta': 50.78264896539114,
'E_theta': 6.201978579936563,
'E_delta': 2.5951045195096842,
'alpha_beta_ratio': 0.1726608027419064,
'peak_frequency': 15.968810916179336,
'spectral_centroid': 15.88220569284006,
'spectral_slope': -10.111986467222136}
```

Accuracy: 0.8150				
Precision: 0.8269				
Recall: 0.8190				
F1 Score: 0.8230				
Confusion Matrix:				
[[77 18]				
[19 86]]				
Classification Report:				
	precision	recall	f1-score	support
0	0.80	0.81	0.81	95
1	0.83	0.82	0.82	105
accuracy			0.81	200
macro avg	0.81	0.81	0.81	200
weighted avg	0.82	0.81	0.82	200

Prototype Stage



Initial Prototype



Final Prototype

Our EEG-operated wheelchair project has been **successfully tested and completed**. We collected data from patients and utilized machine learning to identify and train on different patterns. Each pattern was rigorously tested under various circumstances to enhance accuracy.

As a result, **we have achieved precise control over different actions** including moving forward, backward, turning right, turning left, and stopping. This project demonstrates significant advancements in assistive technology, offering improved mobility and independence for individuals with limited physical capabilities.