

HW III: NEUROMORPHIC SYSTEMS

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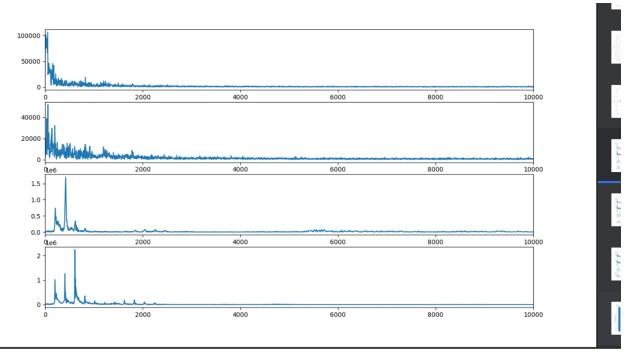
PART 1 CODE

```
import sounddevice as sd
import matplotlib.pyplot as plt
import numpy as np
from scipy.fft import fft
import scipy.signal as signal
import wave
FORMAT = wave.WAVE FORMAT PCM
CHANNELS = 1
DURATION = 1 # Duration of each audio sample in seconds
N SLICES = 4 # Number of time slices to break down the audio signal
    recording = sd.rec(int(DURATION * RATE), samplerate=RATE,
channels=CHANNELS, dtype='int16')
    sd.wait()
    digit recordings.append(recording.flatten())
    slice size = int(len(digit recordings[i]) / N SLICES)
    slices = [digit recordings[i][j*slice size:(j+1)*slice size] for j in
    fig, axs = plt.subplots(N SLICES, figsize=(14, 8))
    fig.suptitle(f"FFT Spectrum of {digit name}")
    for j, slice data in enumerate(slices):
        spectrum = fft(slice data)
        freqs = np.fft.fftfreq(len(spectrum), 1/RATE)
        axs[j].plot(freqs[:len(spectrum)//2],
np.abs(spectrum[:len(spectrum)//2]))
        axs[j].set xlim(0, 10000) # Limit to maximum frequency of 10kHz
    plt.show()
center frequencies = np.arange(750, 8000, 1000) # Center frequencies from
energy levels = np.zeros(len(center frequencies))
```

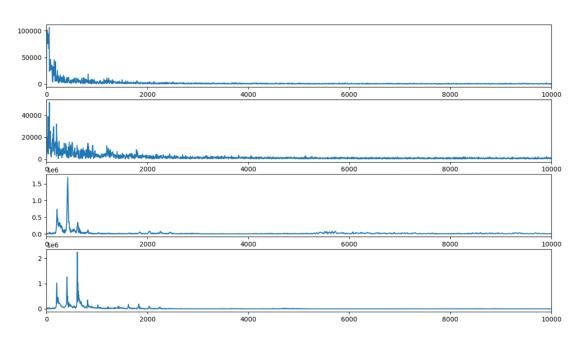
PART ONE AND TWO

```
for i, center freq in enumerate(center frequencies):
    lowcut = center freq - bandwidth / 2
             filtered slice = signal.lfilter(b, a, slice data)
             energy levels[i] += np.sum(np.abs(fft(filtered slice)))
plt.figure(figsize=(10, 6))
plt.bar(center frequencies, energy levels, width=500, align='center')
plt.xlabel('Center Frequency (Hz)')
plt.ylabel('Energy/Power Level')
plt.title('Energy/Power vs Center Frequency')
plt.show()
max power level = np.max(energy levels)
threshold = 0.25 * max_power_level
events matrix = np.zeros((len(center frequencies), N SLICES))
for i, center freq in enumerate (center frequencies):
    for j in range(N_SLICES):
    if energy_levels[i] > threshold:
             events matrix[i, j] = 1
print(events matrix)
```

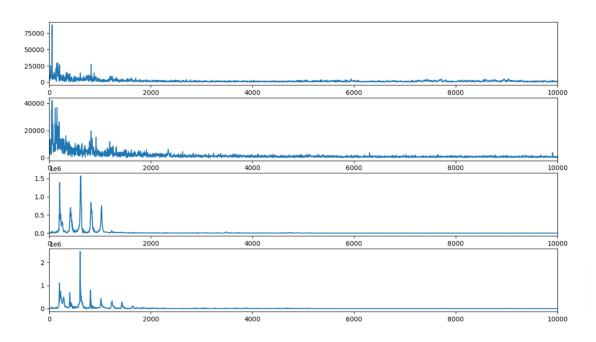
Part 1 OUTPUT -



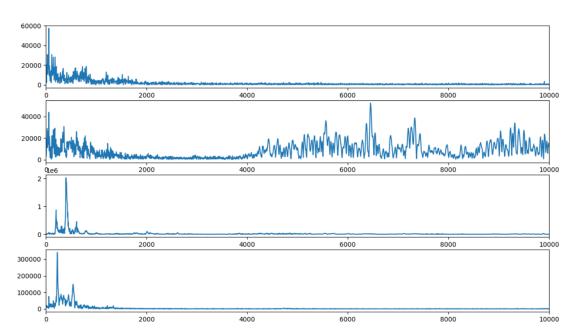
FFT Spectrum of zero

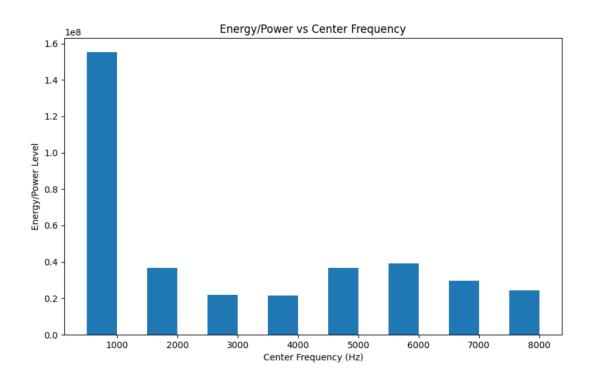


FFT Spectrum of one



FFT Spectrum of two





PART 2 CODE -

```
import numpy as np
import matplotlib.pyplot as plt

def integrate_and_fire_neuron(I_INJECT, V_TH=1, V_RESET=0, G_LEAK=0, CMEM=1e-
12, dt=1e-9, duration=.00001):
    time = np.arange(0, duration, dt)
    V_MEM = np.zeros_like(time)
    spike_count = 0
    V = V_RESET

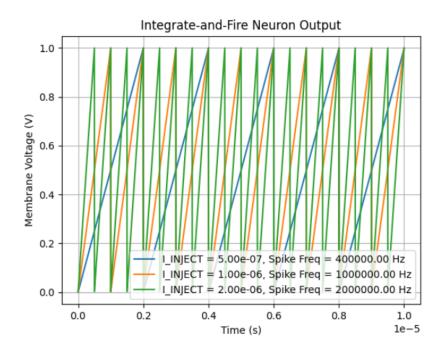
    for i in range(len(time)):
        dV = (I_INJECT - G_LEAK * (V - V_RESET)) / CMEM * dt # Corrected
calculation of dV

    V += dV

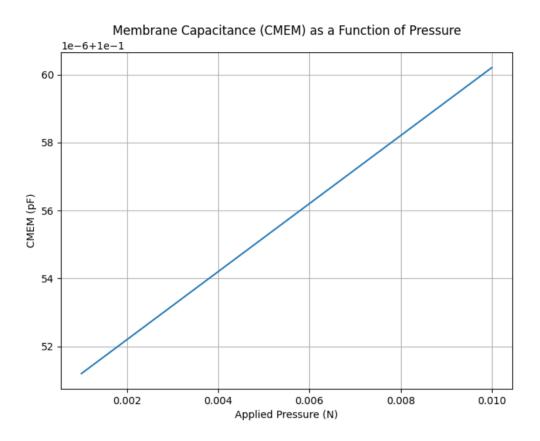
    if V >= V_TH:
        V_MEM[i] = V_TH
        V = V_RESET
        spike_count += 1
    else:
        V_MEM[i] = V
```

```
spike freq = spike count / duration
for I INJECT in I INJECT values:
    time, V_MEM, spike_freq = integrate_and_fire_neuron(I_INJECT)
plt.plot(time, V_MEM, label=f'I_INJECT = {I_INJECT:.2e}, Spike Freq =
{spike freq:.2f} Hz')
plt.xlabel('Time (s)')
plt.ylabel('Membrane Voltage (V)')
plt.title('Integrate-and-Fire Neuron Output')
plt.legend()
plt.grid(True)
plt.show()
def plot spectrum(data, title):
    plt.figure(figsize=(8, 4))
    plt.plot(data)
    plt.title(title)
    plt.show()
def apply_bpf(data, center_freq, bandwidth):
    upper_bound = center_freq + bandwidth / 2
    filtered data = np.where(np.logical and(lower bound <= data, data <=
upper bound), data, 0)
```

PART 2 -



Though difficult to see, we can see the effect of three different currents injected on the behavior of the neuron. Increased current leads to proportionally increased spiking frequency. In an ideal plot, these three injected currents would be displayed on separate plots.



This plot displays C_Mem versus Applied Pressure. C_mem is membrane capacitance of a neuron which refers to the ability of the neuron's membrane to store charge and can determine how rapidly the membrane potential changes when input or signal is input to the neuron. The capacitance plays an important role in action potential propagation and conduction. The relationship between the capacitance of membrane and applied pressure is a linear plot. It can determine pressure sensitivity according to spike rate. This means that pressure is related to spike output of neurons.

PART 3 CODE

```
gleak = 0.0 # Leak conductance (no leak conductance)
def dVdt(V, t, I INJECT):
t = np.linspace(0, 10, 1000) # 10 seconds with 1000 points
plt.figure(figsize=(10, 6))
    spike indices = np.where(V \ge V + (V))
    spike times = t[spike indices]
    spike frequency = len(spike times) / (t[-1] - t[0])
plt.xlabel('Time (s)')
plt.ylabel('Membrane Voltage (V)')
plt.title('Output Membrane Voltage vs Time')
plt.legend()
plt.grid(True)
plt.show()
```

Part 3 – Code

```
import numpy as np
import matplotlib.pyplot as plt

# Number of stages in the ring oscillator
num_stages = 6

# Time array (assuming 1 second duration)
t = np.linspace(0, 1, 1000)

# Function to generate sine wave outputs for each stage
def generate_sine_wave(phase):
    return np.sin(2 * np.pi * (t - phase))

# Part 1: Plot the sine wave outputs of the 6 stages
plt.figure(figsize=(10, 6))
for i in range(num_stages):
    plt.plot(t, generate_sine_wave(i / num_stages), label=f'Stage {i + 1}')

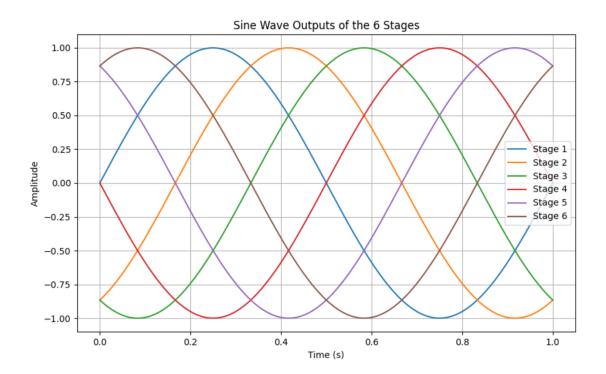
plt.title('Sine Wave Outputs of the 6 Stages')
plt.ylabel('Time (s)')
plt.ylabel('Amplitude')
plt.legend()
plt.grid(True)
plt.show()

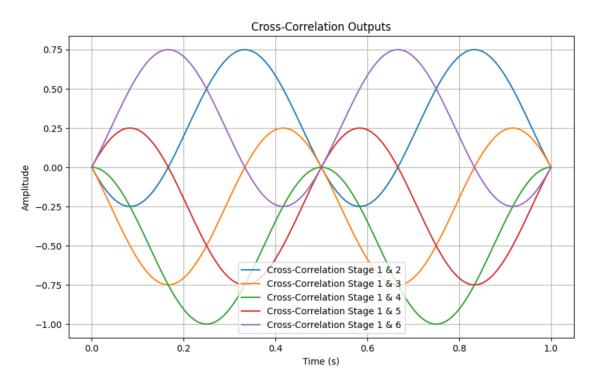
# Part 2: Plot the cross-correlation outputs
```

PART ONE AND TWO

```
plt.figure(figsize=(10, 6))
for i in range(1, num stages):
    cross corr = np.multiply(generate sine wave(0), generate sine wave(i /
num stages))
    plt.plot(t, cross corr, label=f'Cross-Correlation Stage 1 & {i + 1}')
plt.title('Cross-Correlation Outputs')
plt.xlabel('Time (s)')
plt.ylabel('Amplitude')
plt.legend()
plt.grid(True)
plt.show()
threshold = 0.1
unique codes = []
    cross corr = np.multiply(generate sine wave(0), generate sine wave(i /
   code = [1 if val > threshold else 0 for val in cross corr]
    unique codes.append(code)
print("Unique Codes for Each Phase:")
for i, code in enumerate(unique codes):
```

Part 3 - Output





3.3: /Users/jaspreetsingh/Documents/NeuromorphicHW3/.venv/bin/python /Users/jaspreetsingh/Documents/NeuromorphicHW3/hw3p3.py Unique Codes for Each Phase:

Process finished with exit code 0

PART ONE AND TWO

The codes capture the cross-correlation outputs and the applied threshold by providing a simplified digital representation of the degree of similarity or dissimilarity of the cross-correlation outputs in relation to the specified threshold level. The binary code will show the cross-correlation output as above (1) or below (0) which would allow for differentiation between phases when given the same threshold.