

# Meeting 6

10/14/21

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# Deliverables

- Explain Fourier Transform
- Expand application of welch function to all 64 electrodes

# Methodology and Learnings

How did I do it?

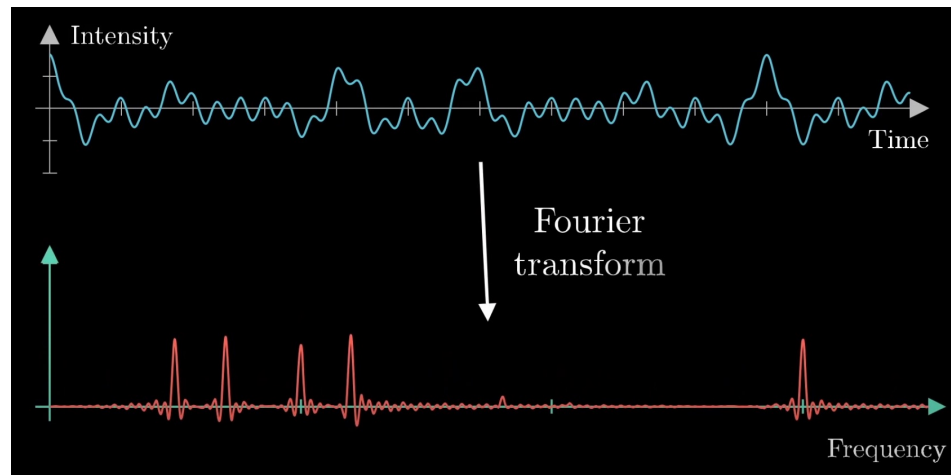
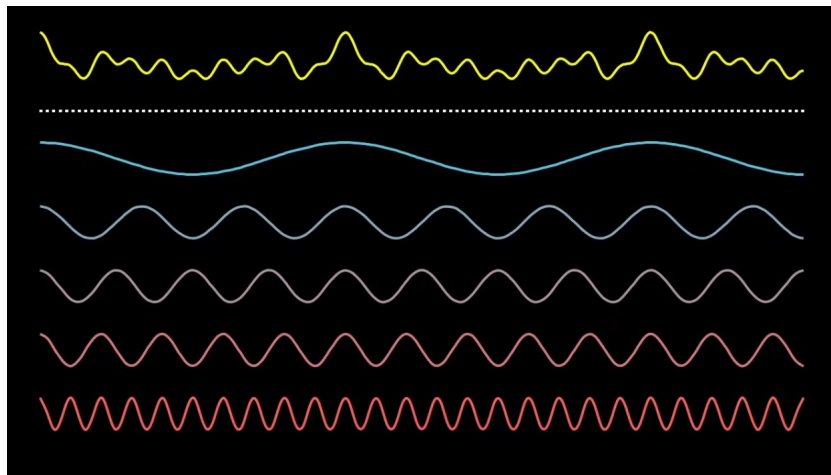
- Online research

What did I learn on the way?

- Fourier transform

# Fourier transform

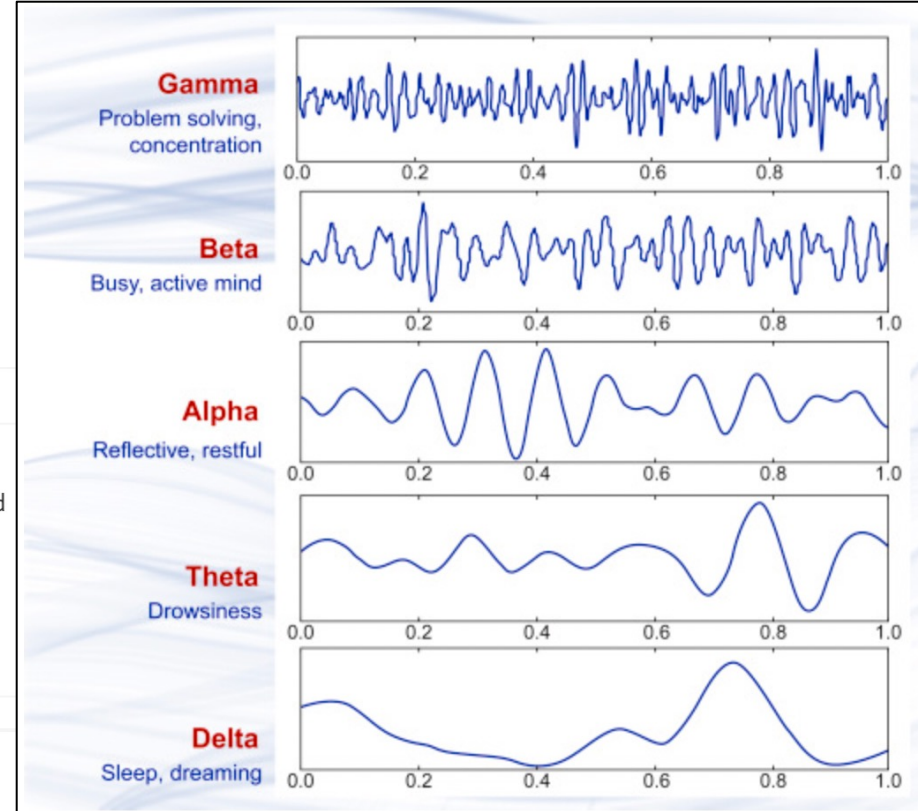
- Another way to represent a waveform
- Multiple applications
- Useful with EEG signals



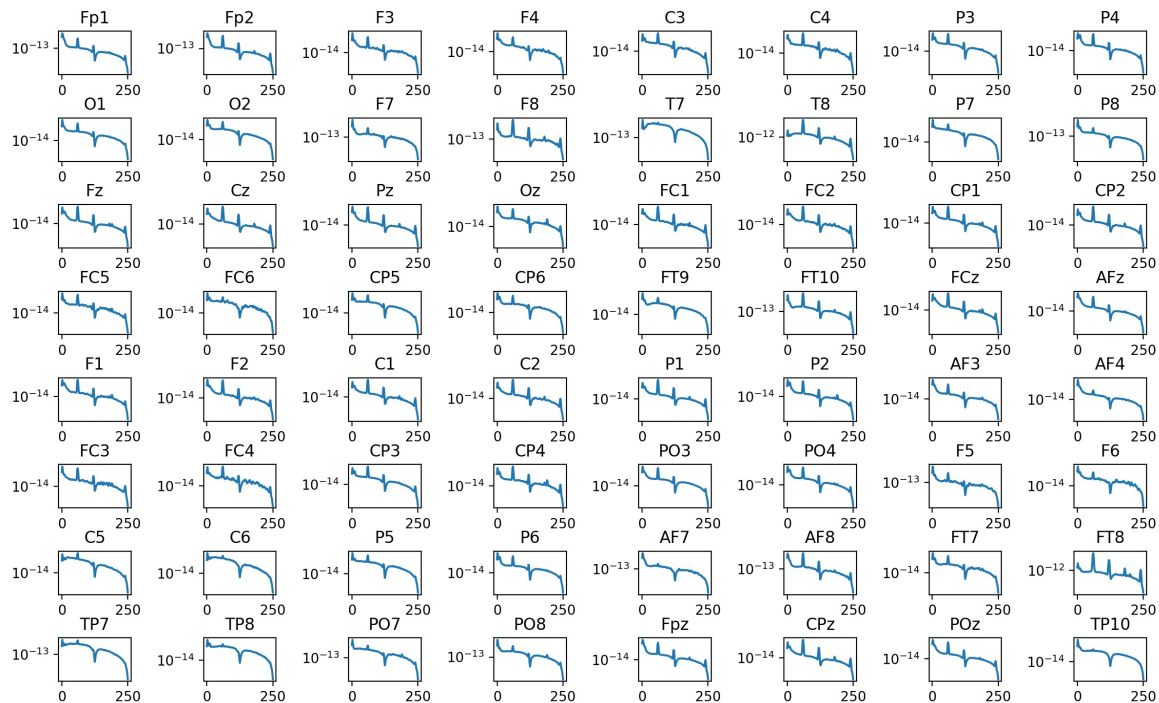
# How it relates

- Interpret data easier
- Patterns become visible

Frequency band	Frequency	Brain states
Gamma ( $\gamma$ )	$>35$ Hz	Concentration
Beta ( $\beta$ )	12–35 Hz	Anxiety dominant, active, external attention, relaxed
Alpha ( $\alpha$ )	8–12 Hz	Very relaxed, passive attention
Theta ( $\theta$ )	4–8 Hz	Deeply relaxed, inward focused
Delta ( $\delta$ )	0.5–4 Hz	Sleep



# Results



## Results – cont.

```
# Read data
data = read_eeg("./2020_06_04_T05_U00T_EEG01.vhdr")

# create dataframe for 64 signals
df = data.iloc[:, 0:64]

# plot data
row = 8
col = 8
count = 1

# dimensions of plot
fig = plt.figure(figsize=(8, 8))
fig.subplots_adjust(hspace=.9, wspace=.9)

# Traverse the dataframe and create a plot for each EEG signal
for i in df.columns:
    freq_array, psd = signal.welch(df[i], fs=500) # store frequency and power spectrum of signals
    plt.subplot(row, col, count) # create subplot
    plt.semilogy(freq_array, psd) # plot the semilogy graph
    plt.title(i) # title of graph
    count = count + 1 # increment plot counter
plt.show()
```