

Time Frequency Analysis

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Outline

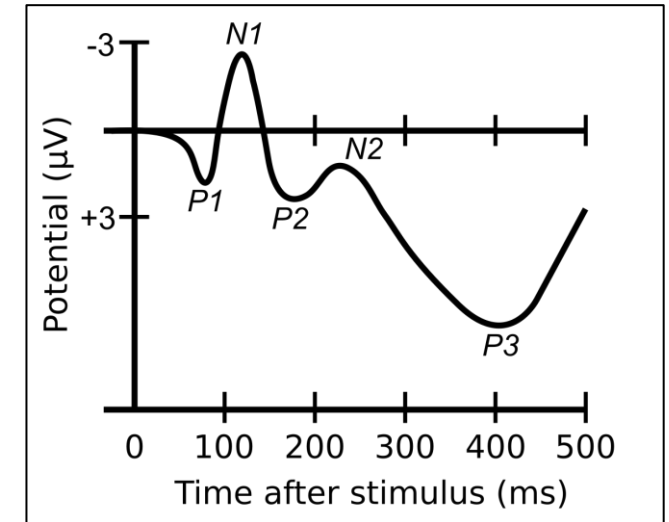
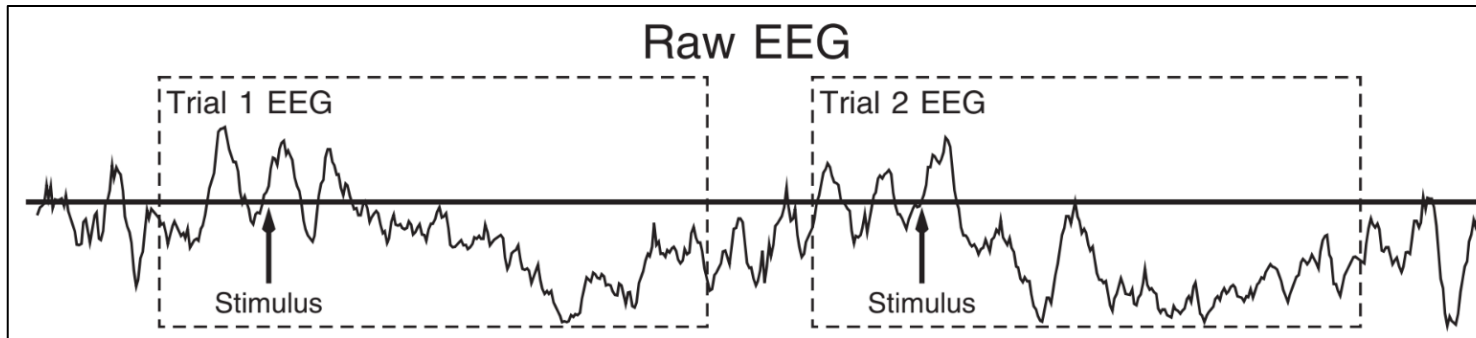
1. There's more in the data than ERPs
2. What are brain oscillations?
3. How to compute them
 1. Properties of a wave
 2. Fourier transform
 3. Practical tips
 4. Visualization in EEG
4. Outlook: more advanced / other methods

**Not covered: python
implementation**

There's more in the data than ERPs

Basic idea of ERPs:

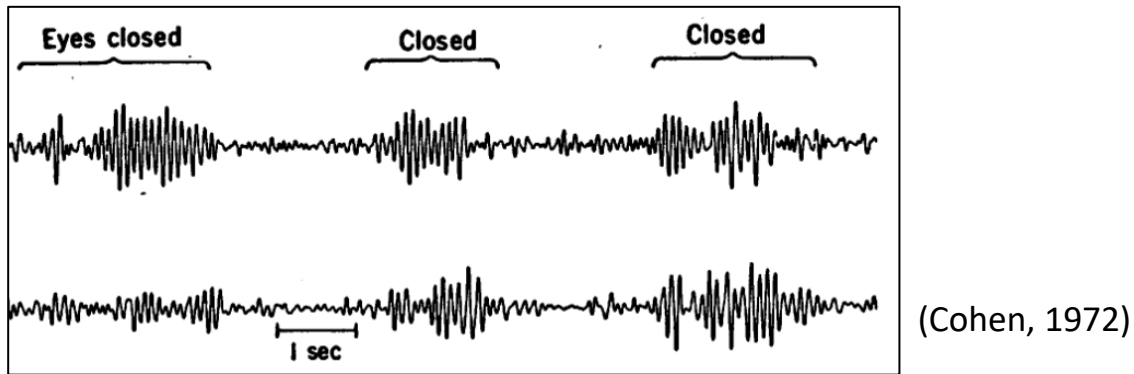
- Brain response to a discrete, isolated event
→ **time-locked and averaged**



A lot of information gets lost by averaging!

There's more in the data than ERPs

A lot of information gets lost by averaging!



Other aspects that ERPs can't capture:

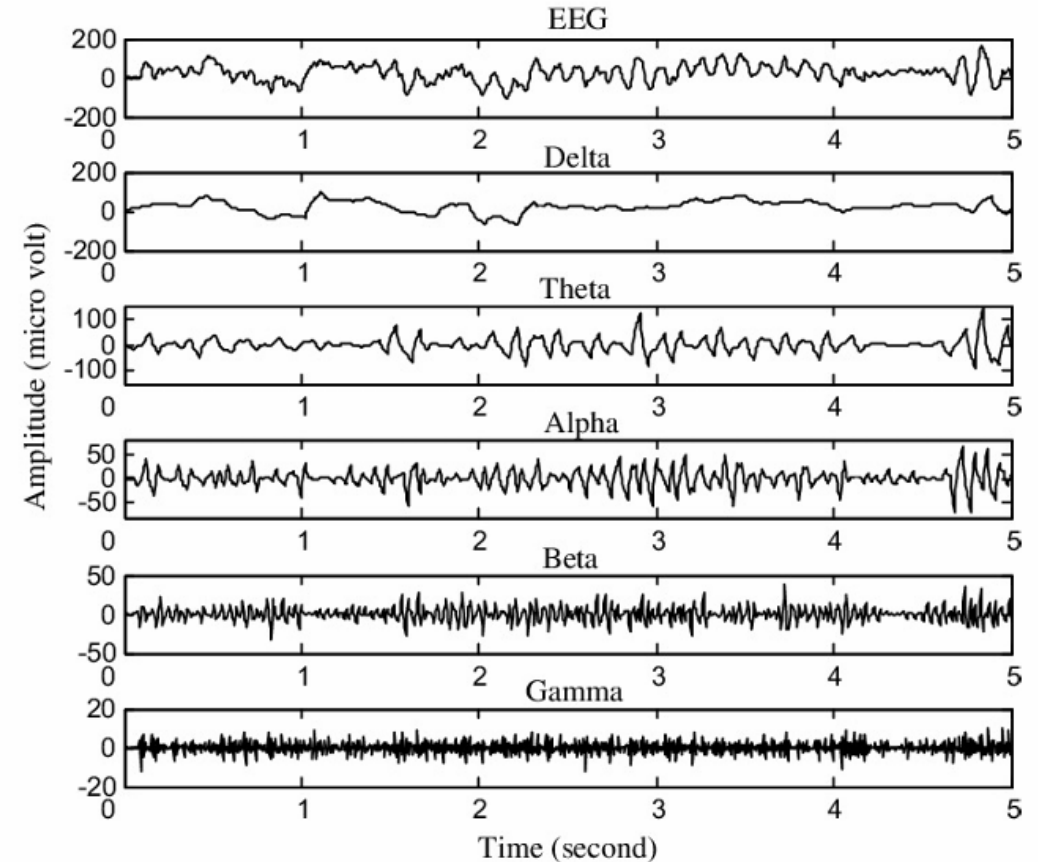
- Natural stimuli like speech
- Intrinsic brain functions (not triggered by event)

Brain Oscillations

- rhythmic / repetitive patterns of brain activity
- oscillations of different frequencies overlaying in our signal

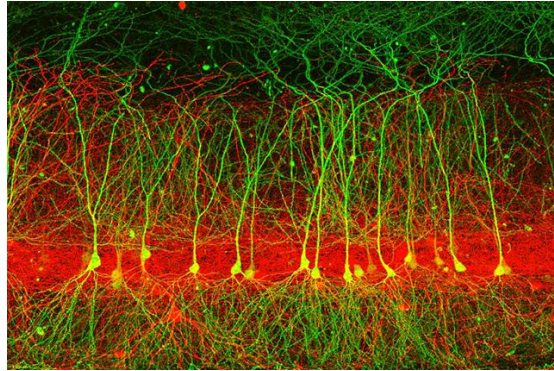
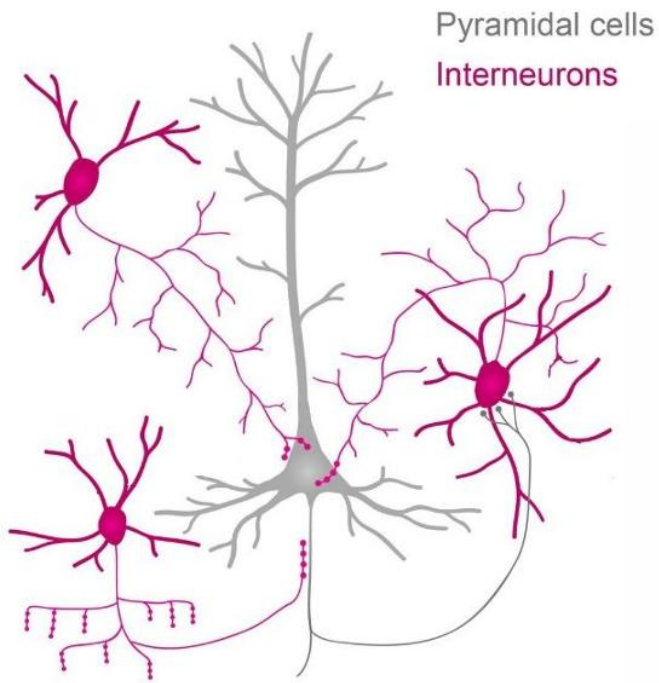
Why is there oscillating activity in the brain?

- Coordination of activity in neural ensembles → synchronous firing
- Communication between brain regions
- Feature binding

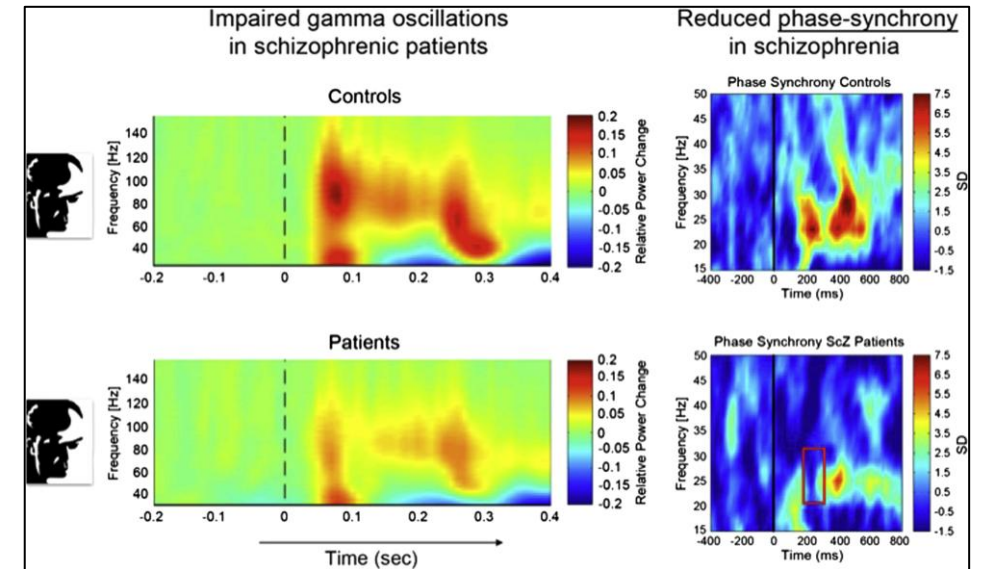


Brain Oscillations

- Coordination of activity in neural ensembles → synchronous firing



→ recurrent,
synchronized activity
in gamma range

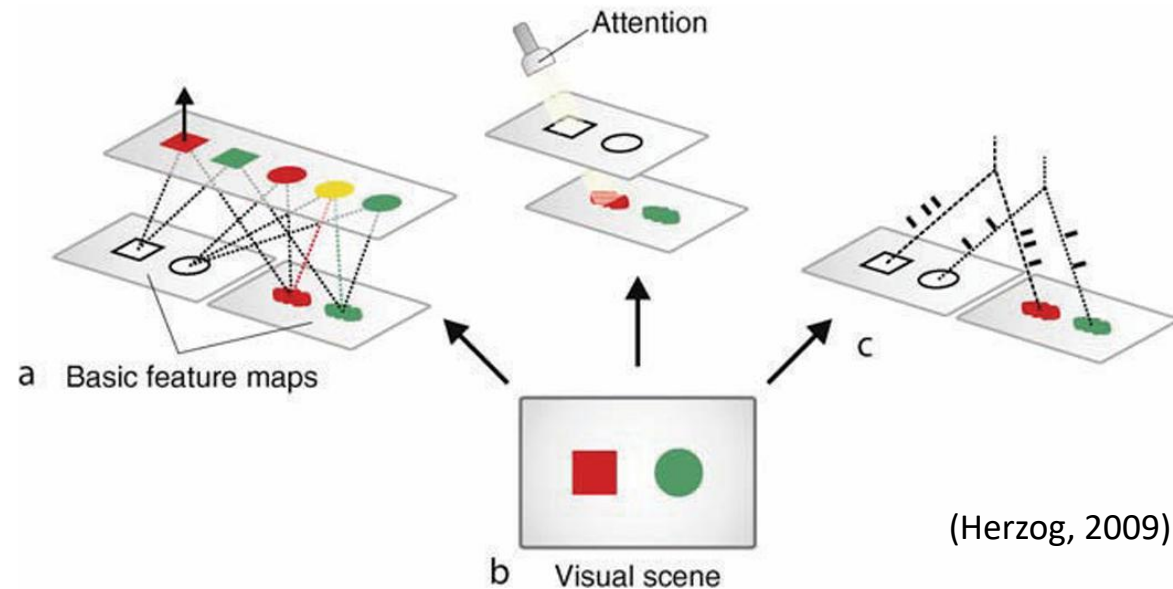


(Buszaki et al., 2013)

Brain Oscillations

Why is there oscillating activity in the brain?

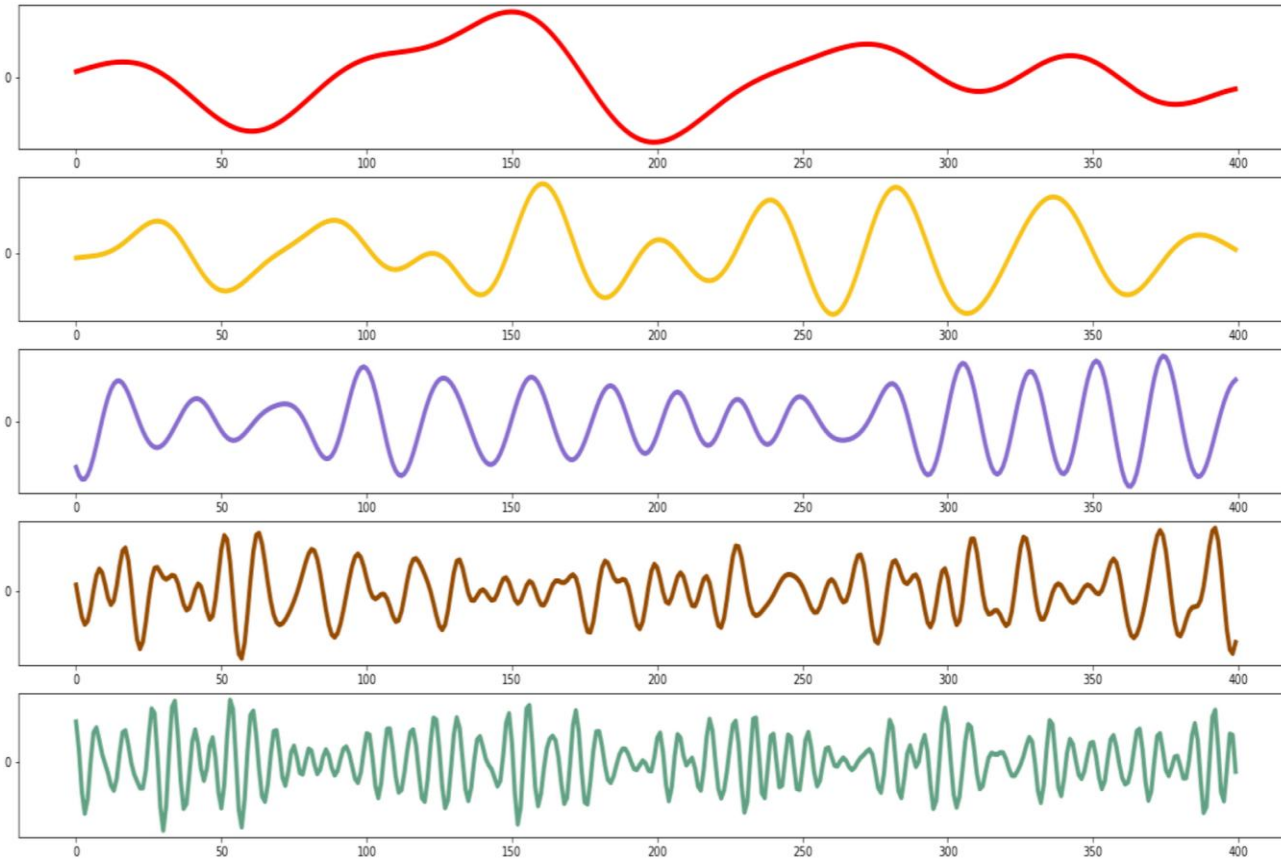
- Feature Binding



Oscillations could be **possible solution to binding problem**:

- Neurons coding for features belonging to the same object fire in synchrony

Brain Oscillations



Delta (< 4 Hz): deep sleep

Theta (4-8 Hz): dreaming, working memory

Alpha (8-12 Hz): relaxed wakefulness

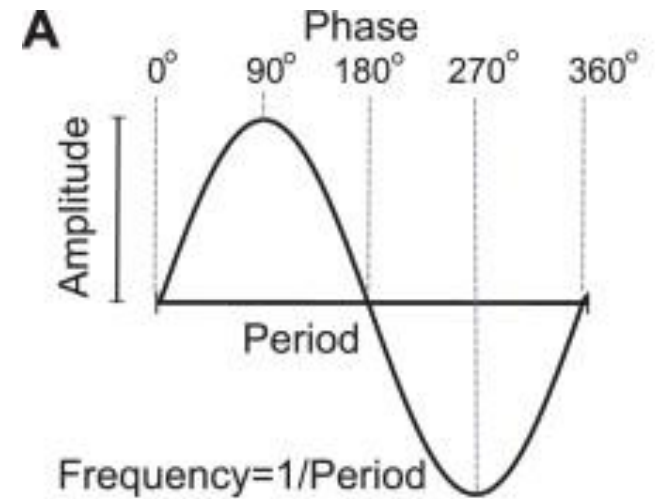
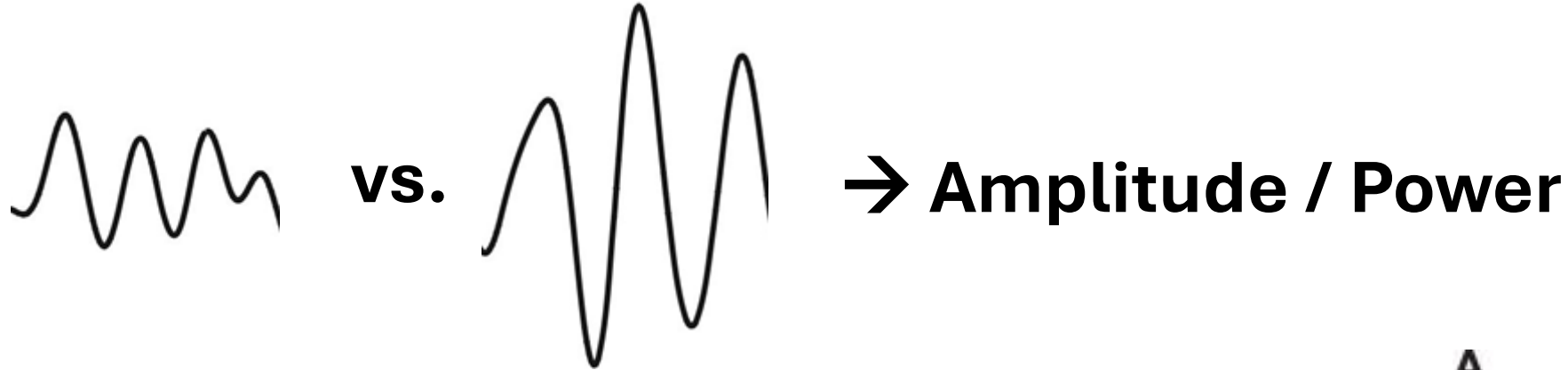
Beta (13-30 Hz): active concentration

Gamma (30-90 Hz): sensory input

(Pandey et al., 2022)

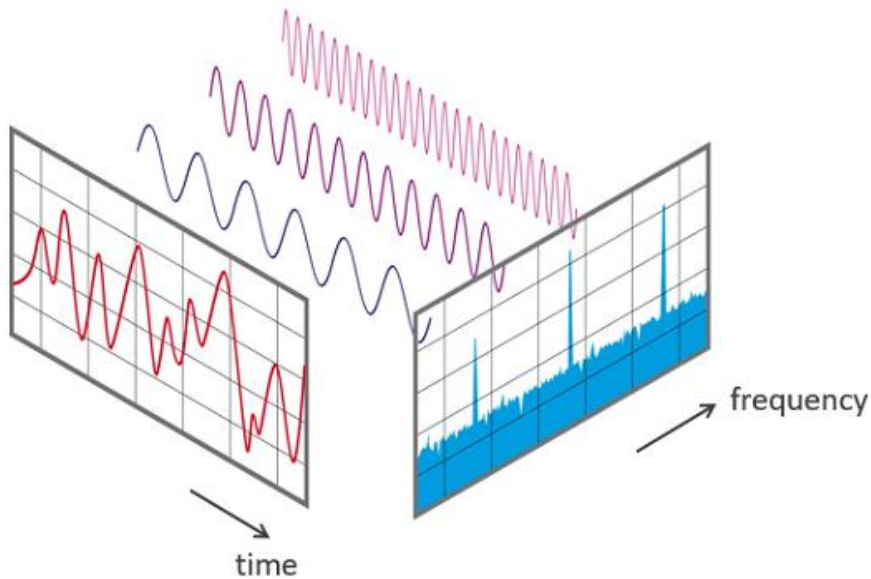
Time-frequency analysis

How can one describe a wave?



Fourier Transform

Idea: You can decompose **every continuous signal** into sine waves of varying frequencies, amplitudes and phases



Calculate power for each frequency

→ Frequency domain

→ Loose time information

→ You can slide a window through your data to get some time information

Wavelet convolution

To extract frequency-specific information from EEG data the data must be **convolved with a sine wave**

C) **one cycle** maximizes temporal precision but less frequency precision

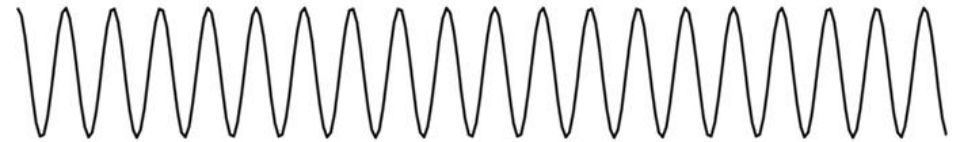
D) **boxcar tapering**: decreased temporal specificity and potential artifacts from sharp edges

E) Gaussian tapering (= **Morlet wavelet**): balance between temporal and frequency precision

A) EEG data



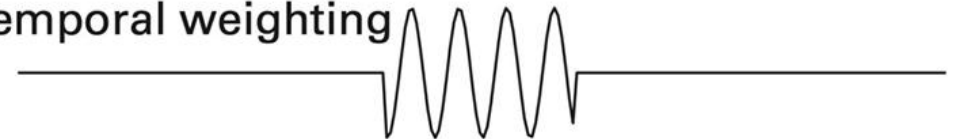
B) No temporal weighting (Fourier transform)



C) Strong temporal weighting



D) Boxcar temporal weighting

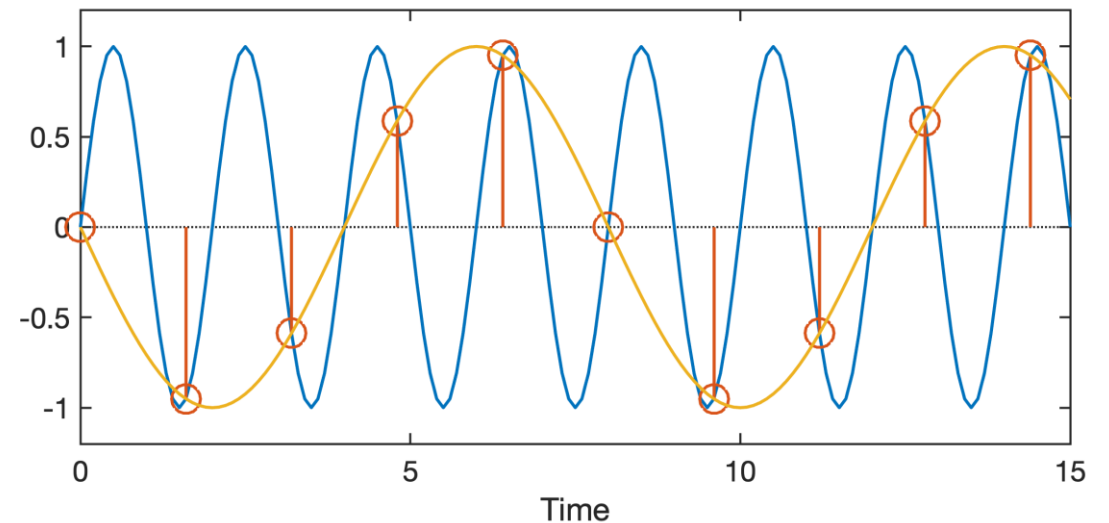


E) Gaussian temporal weighting



Basic Rules

1. You can only analyze frequencies for which you have several cycles in your epoch (e.g. 1s epoch \rightarrow not less than 4 Hz).
2. Don't look at frequencies above the Nyquist frequency (= half your sample rate): sampling rate of 500 Hz \rightarrow up to 250 Hz

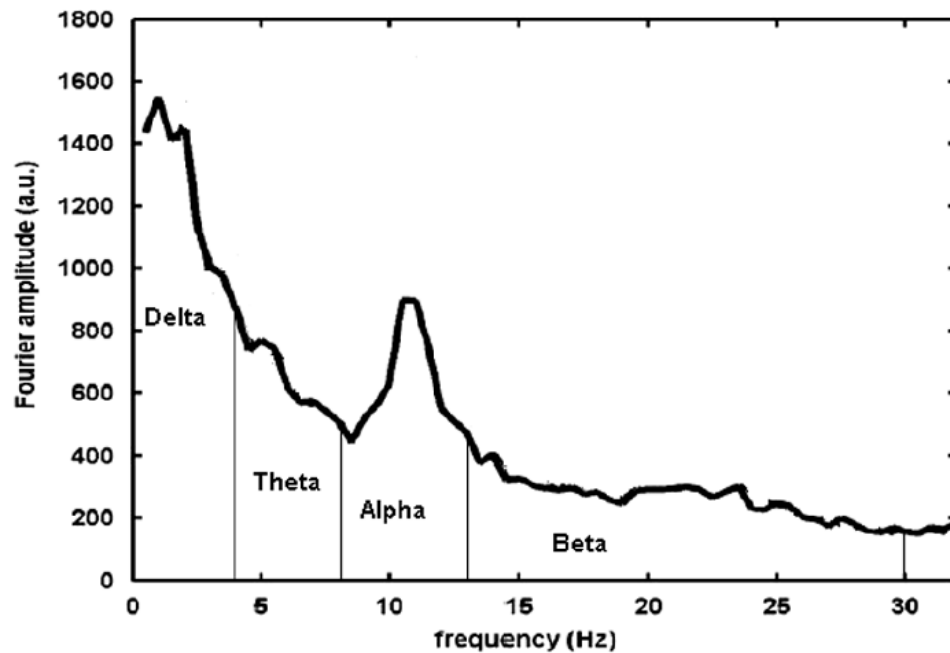


Basic Rules

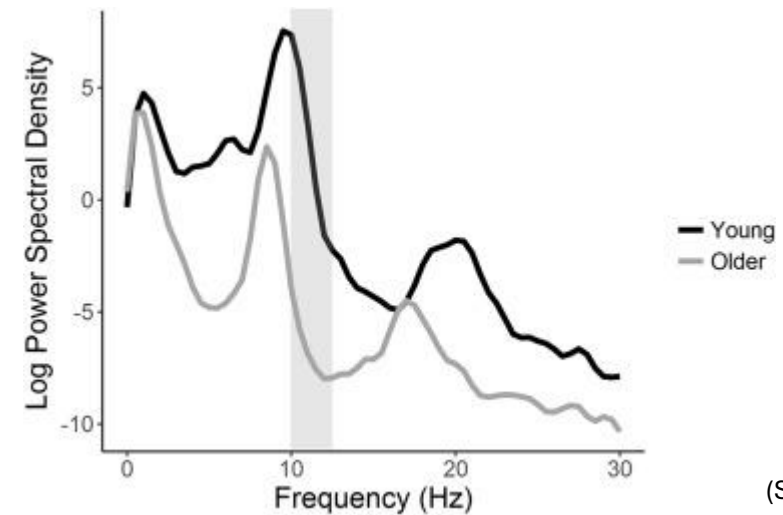
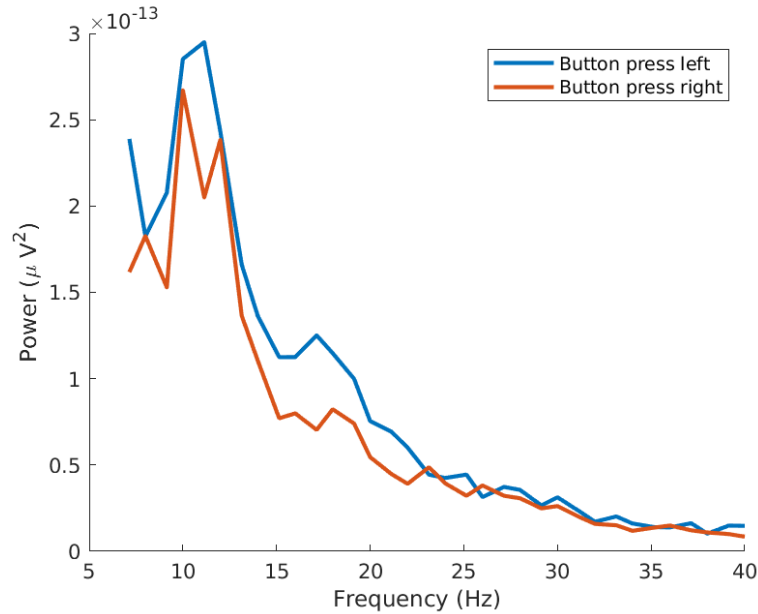
1. You can only analyze frequencies for which you have several cycles in your epoch (e.g. 1s epoch \rightarrow not less than 4 Hz).
2. Don't look at frequencies above the Nyquist frequency (= half your sample rate): sampling rate of 500 Hz \rightarrow up to 250 Hz
3. Frequencies that are very close to each other will provide similar results (because of time-frequency precision trade-offs)

Visualization in EEG

Frequency Domain (PSD plot)
(no temporal resolution)



(Namazi & Kulish, 2012)

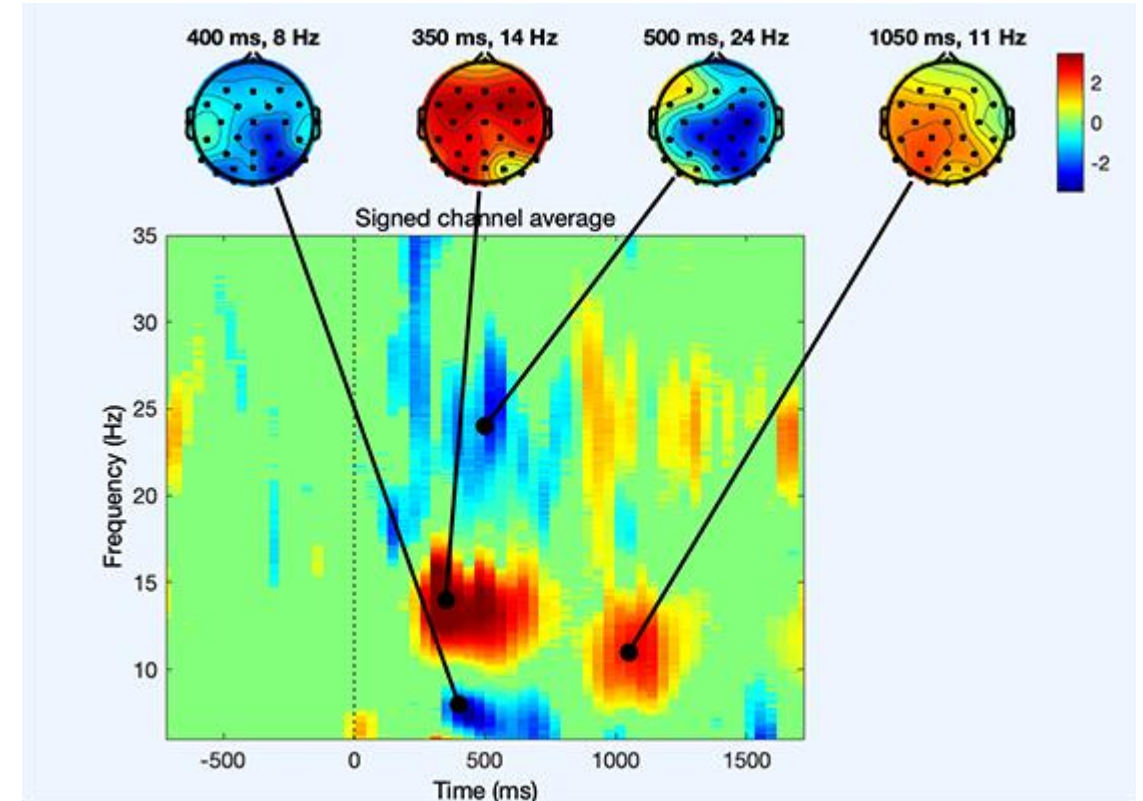
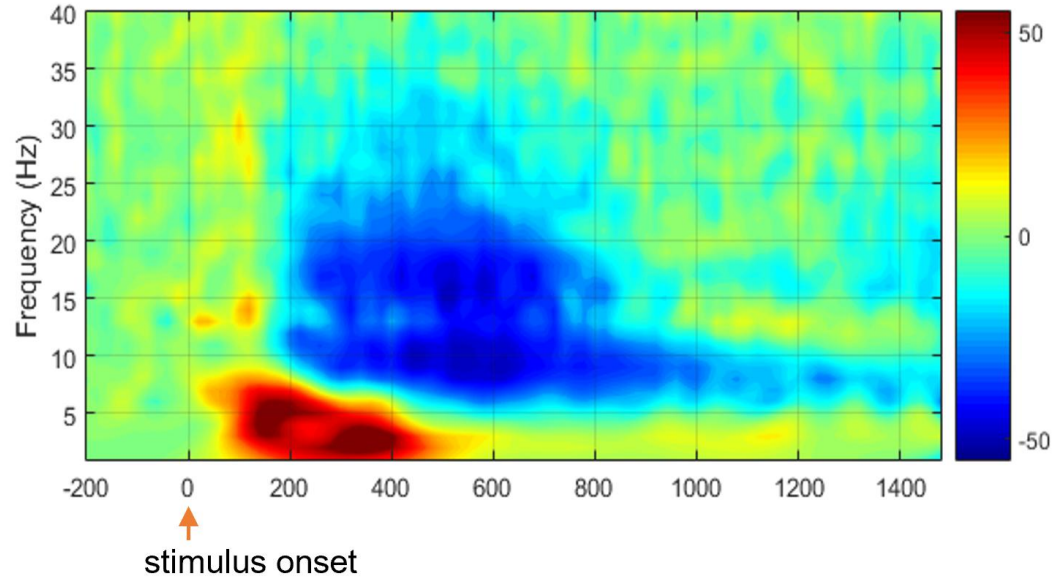


(Sally et al., 2018)

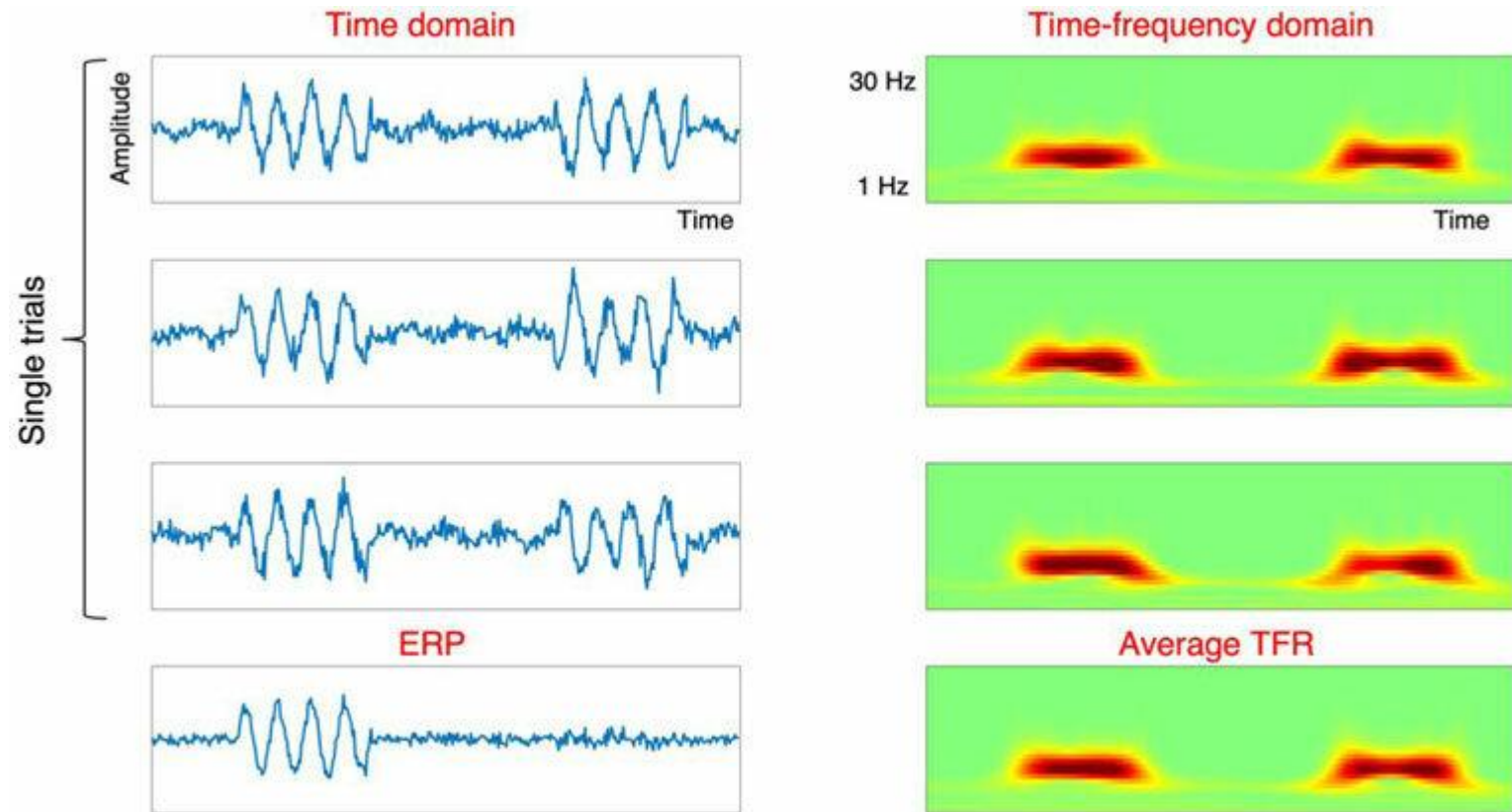
Power is often Log-scaled!

Visualization in EEG

Frequency and Time



Evoked vs. Induced Oscillations



(Rossi et al., 2022)

Caveat

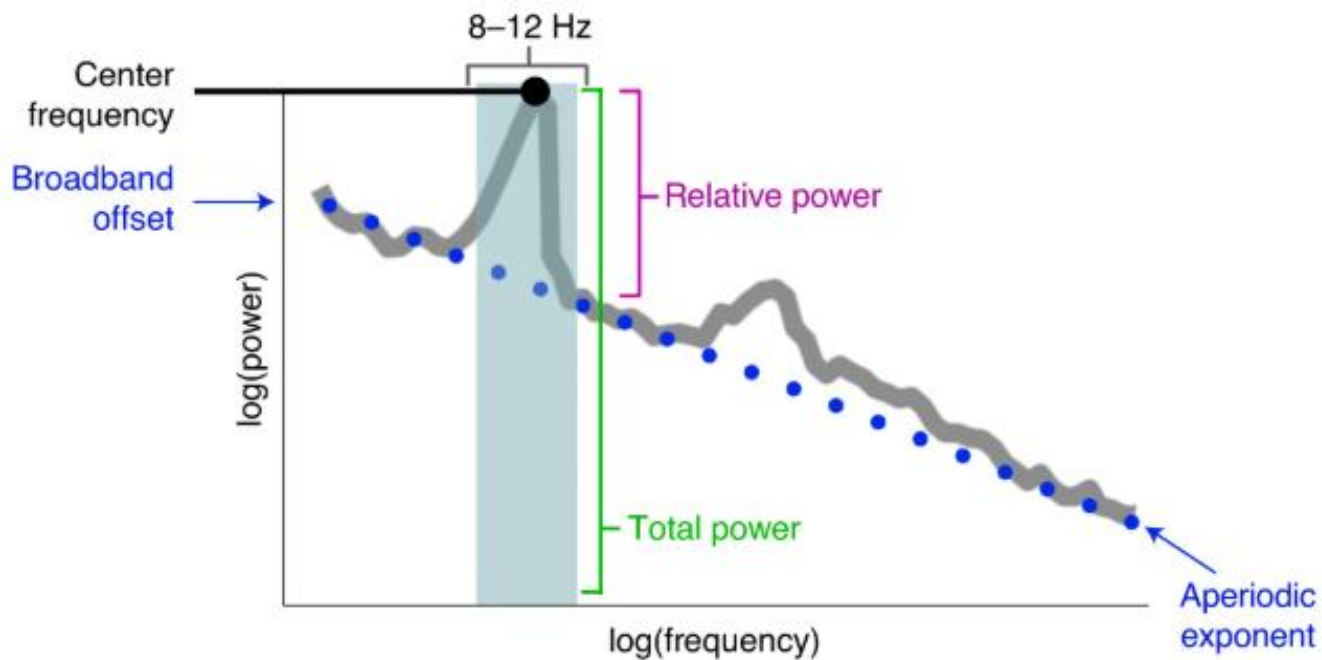
Power at a given frequency does not necessarily mean that the brain was oscillating at that frequency...

Think of the definition of the Fourier principle!

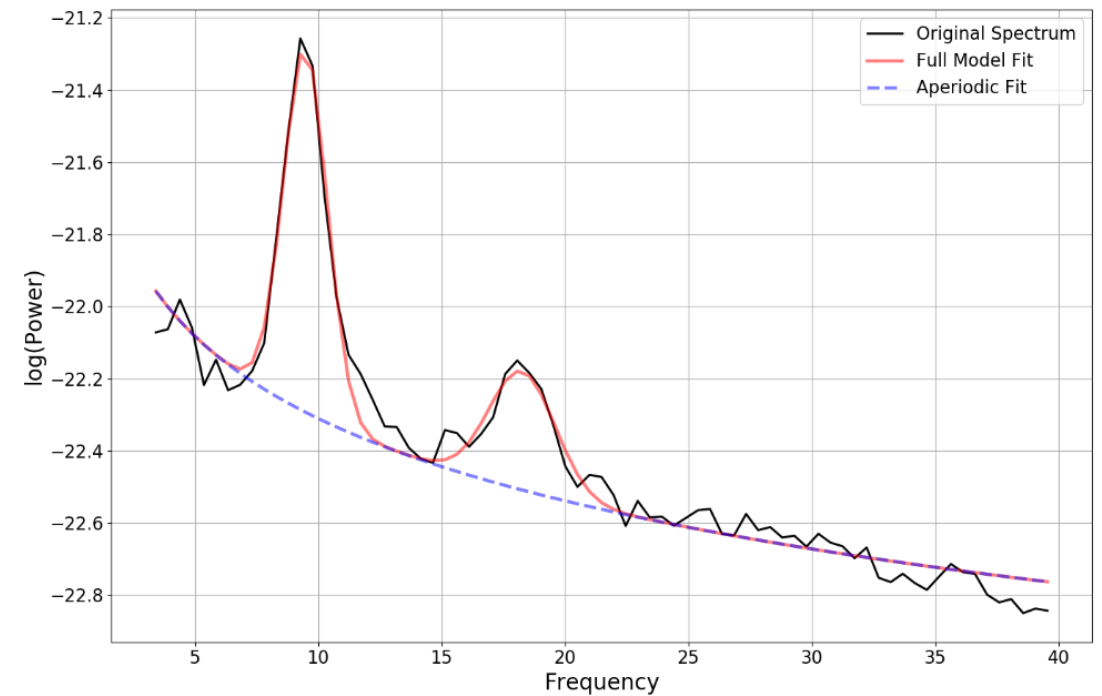
- Lower frequencies: ERP-like brain processes
 - Higher frequencies: random neural noise /background activity
 - Artefacts like blinks, heart beat, movement, line noise lead to power in specific bands
- Clean data, no systematic differences between conditions!

Additional analysis methods

- Separating oscillatory (periodic) from non-oscillatory (aperiodic) power

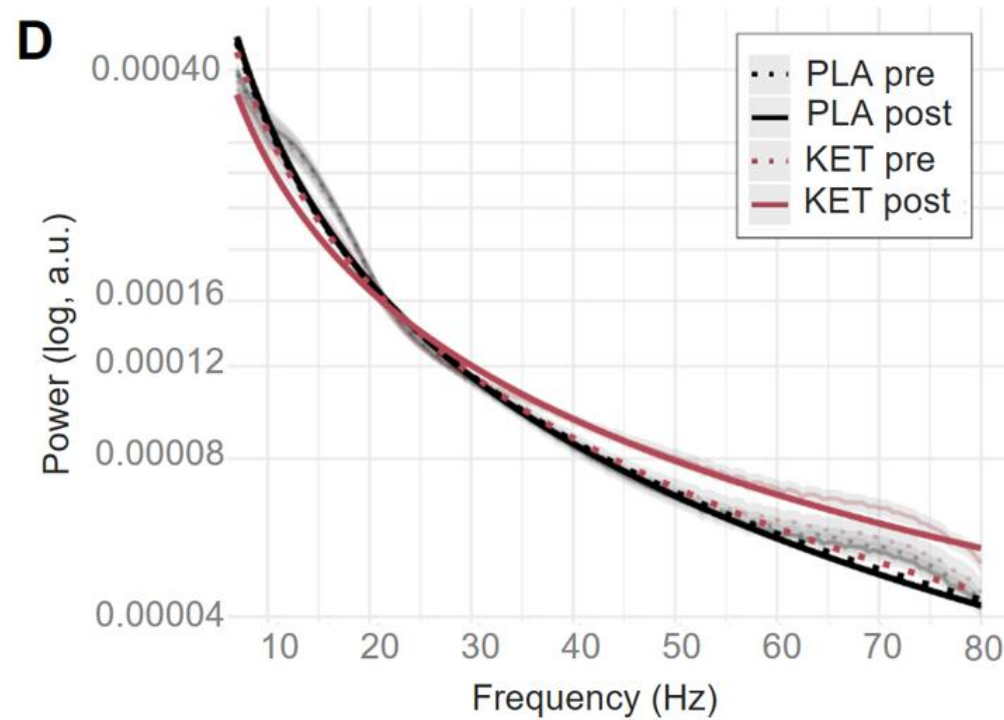


(Donoghue et al., 2020)



Additional analysis methods

- Separating oscillatory (periodic) from non-oscillatory (aperiodic) power



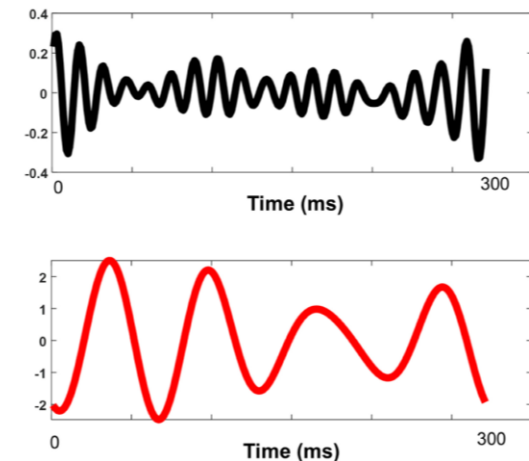
Additional analysis methods

- Cross-Frequency-Coupling
- activity of neuronal assemblies, each supporting a specific frequency band, influences or modulates the activity of others
- Phase-amplitude coupling: amplitude of faster oscillations is coupled to phase of slower oscillations

Gamma amplitude modulated by beta phase



(Hodnik et al. 2024)



What we've covered



Why time-frequency analysis complements ERP analysis



Potential physiological mechanisms of oscillations



How to compute time-frequency analysis



Outlook on more advanced / other methods

Any Questions?

Partly based on
“Analyzing Neural Time Series Data: Theory and Practice”

Mike X Cohen
The MIT Press, 2014