

# Spectral Time Series Analysis

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## 1 Introduction

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# Oscillations in nature

Today we will characterize oscillations in neural time series. But oscillations are everywhere in nature. For example:

**circadian rhythm** the 24 hour cycle that governs sleep, hormone release, body temperature and other physiological processes.

**menstrual cycle** a roughly 28-day hormonal cycle regulating ovulation and fertility in females, driven by oscillations in estrogen and progesterone levels.

**ultradian rhythms of physical performance** these are cycles shorter than a day that governs periods of focus and fatigue. Athletes often perform better at specific times of the day due to oscillatory patterns in physical strength and endurance.

**animal migration** many animal migrations are governed by oscillatory patterns linked to seasonal changes (e.g., birds migrating during spring and autumn).

# Oscillations in nature

**population dynamics** predator-prey systems exhibit oscillatory behavior.

When prey populations rise, predator populations grow, leading to a decline in prey, and then a subsequent decline in predators.

**lunar cycles** the moon's orbit around the Earth (roughly 28 days) governs the tides, influencing ecosystems like coral spawning and animal behaviors.

# Oscillations in the brain

Oscillations, measured in different frequency bands (e.g., delta, theta, alpha, beta, gamma), play a crucial role in various cognitive functions, including:

- Synchronization and Communication:

**Binding** Oscillations enable the brain to synchronize the activity of neurons across different regions, allowing them to work together as a cohesive unit. This is particularly important for tasks like perception, where information from different senses needs to be integrated.

**Information Transfer** Oscillations can act as a "clock" to coordinate the timing of neuronal firing, facilitating efficient communication between brain areas.

# Oscillations in the brain

- Cognitive Functions:

**Attention** Different frequency bands are associated with different attentional states. For example, alpha waves are prominent during relaxed wakefulness, while beta waves are associated with focused attention.

**Memory** Oscillations have been implicated in memory processes, particularly in the consolidation of information from short-term to long-term memory.

- Sensory Processing:

**Sensory Perception** Oscillations help to filter out irrelevant sensory information and enhance the processing of relevant stimuli.

**Motor Control** Oscillations play a role in coordinating the timing of muscle movements, ensuring smooth and efficient motor actions.

# Oscillations in the brain

- Sleep:

**Defining Sleep Stages** Different sleep stages (NREM and REM) are characterized by distinct patterns of brain oscillations.

**NREM Sleep** Dominated by slow waves (delta) and sleep spindles.

**REM Sleep** Characterized by fast, low-amplitude waves resembling wakefulness.

**Sleep Functions Memory Consolidation** Slow waves and sleep spindles are crucial for strengthening memories.

**Synaptic Plasticity** Sleep oscillations contribute to the strengthening and weakening of connections between neurons.

**Brain Rest and Recovery** Slow waves may help the brain recover from the demands of wakefulness.



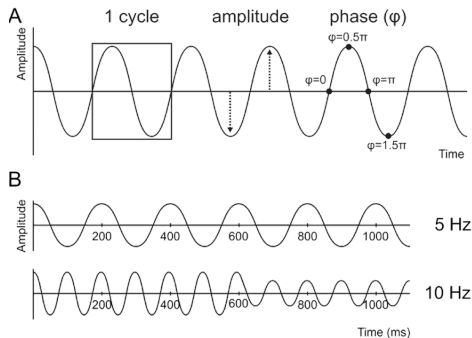
# Describing oscillations

$$x(t) = A \cos(2\pi f t + \phi)$$

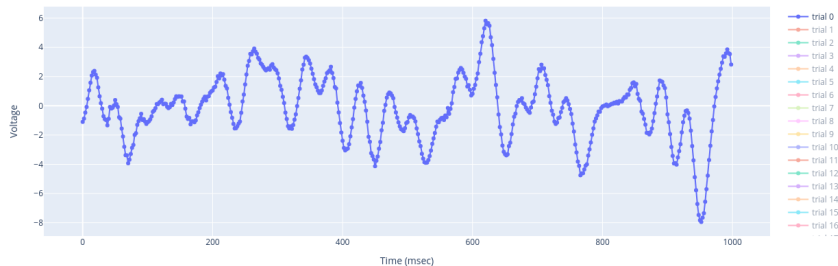
$A$  amplitude ( $\mu V$ )

$f$  frequency (Hertz – 1/sec)

$\phi$  phase (radians)



# Example EEG time series



Liu et al. (2024)

We will use complex exponentials to represent oscillations.

$$e^{j\Omega t} = \cos(\Omega t) + j \sin(\Omega t) \quad \text{with } j^2 = -1$$

# Fourier transforms for deterministic signals (1/4)

Name Fourier integral

Time domain type continuous aperiodic

Frequency domain type continuous aperiodic

Formulas

$$x(j\Omega) = \int_{-\infty}^{\infty} x(t) e^{-j\Omega t} dt$$

$$x(t) = \frac{1}{2\pi} \int_{-\infty}^{\infty} X(j\Omega) e^{j\Omega t} d\Omega$$

Existence (sufficient condition)

$$\int_{-\infty}^{\infty} |x(t)| dt < \infty$$

# Fourier transforms for deterministic signals (2/4)

Name Fourier series

Time domain type continuous periodic

Frequency domain type discrete infinite

Formulas

$$x[k] = \int_{-T/2}^{T/2} x(t) e^{-j \frac{2\pi}{T} kt} dt$$

$$x(t) = \sum_{k=-\infty}^{\infty} X[k] e^{j \frac{2\pi}{T} kt}$$

# Fourier transforms for deterministic signals (3/4)

Name Discrete Fourier transform

Time domain type discrete infinite

Frequency domain type continuous periodic

Formulas

$$x(j\omega) = \sum_{n=-\infty}^{\infty} x[n]e^{-j\omega n}$$
$$x[n] = \frac{1}{2\pi} \int_{-\pi}^{\pi} X(j\omega)e^{j\omega n} d\omega$$

# Fourier transforms for deterministic signals (4/4)

Name Discrete Time Fourier transform

Time domain type discrete finite

Frequency domain type discrete finite

Formulas

$$x[k] = \sum_{n=1}^N x[n] e^{-j \frac{2\pi}{N} nk}$$

$$x[n] = \sum_{k=1}^N x[k] e^{j \frac{2\pi}{N} nk}$$

## Definition 1 (spectral density function)

A function  $f$  is the **spectral density** of a stationary time series  $\{X_t\}$  with autocovariance function  $\gamma(\cdot)$  if

- i)  $f(\lambda) \geq 0$  for all  $\lambda \in (-\pi, \pi]$ , and
- ii)  $\gamma(h) = \int_{-\pi}^{\pi} e^{ih\lambda} f(\lambda) d\lambda$ , for all integers  $h$ .

Note:  $f(\lambda)$  gives the power of a random process in the frequency range  $[\lambda, \lambda + \delta\lambda]$ .



Liu, Q., Jia, S., Tu, N., Zhao, T., Lyu, Q., Liu, Y., Song, X., Wang, S., Zhang, W., Xiong, F., et al. (2024). Open access eeg dataset of repeated measurements from a single subject for microstate analysis. *Scientific Data*, 11(1):379.